



## Accelerating Technology Development: Post Combustion Capture Sorbents

**David C. Miller and James C. Fisher II**

National Energy Technology Laboratory

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U.S. DEPARTMENT OF

**ENERGY**

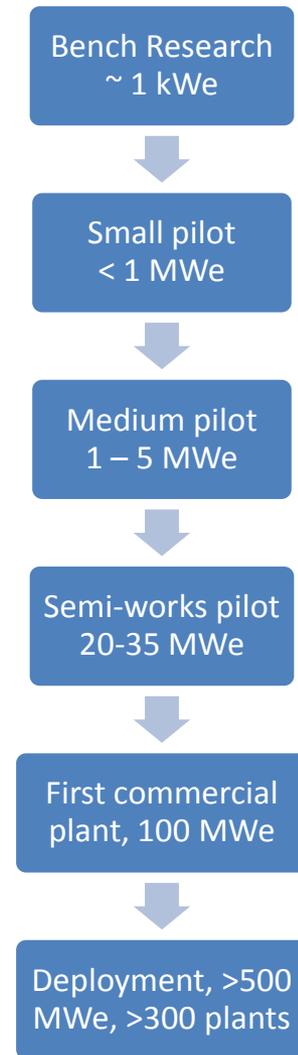
National Energy  
Technology  
Laboratory

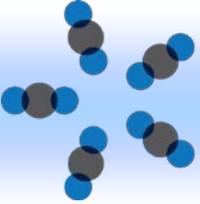
# Outline

- **Carbon Capture Challenge**
- **Computational Tools to Accelerate Technology Development**
- **Experimental Carbon Capture Research @ NETL**
- **Multiscale Model Development, Simulation & Optimization**
- **Experimental Validation**
- **Conclusions**

# Carbon Capture Challenge

- The traditional pathway from discovery to commercialization of energy technologies can be quite long, i.e., 20-30 years
- New approaches are needed for taking carbon capture concepts from lab to power plant, quickly, and at low cost and risk
- Science-based simulations will accelerate the development of carbon capture technology, from discovery through deployment

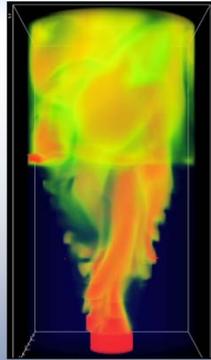
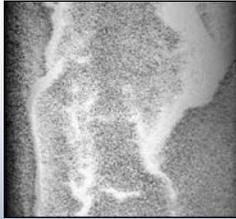
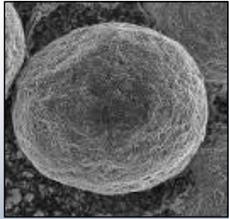




# CCSI

Carbon Capture Simulation Initiative

# For Accelerating Technology Development



Identify promising concepts



Reduce the time for design & troubleshooting



Quantify the technical risk, to enable reaching larger scales, earlier



Stabilize the cost during commercial deployment

## National Labs



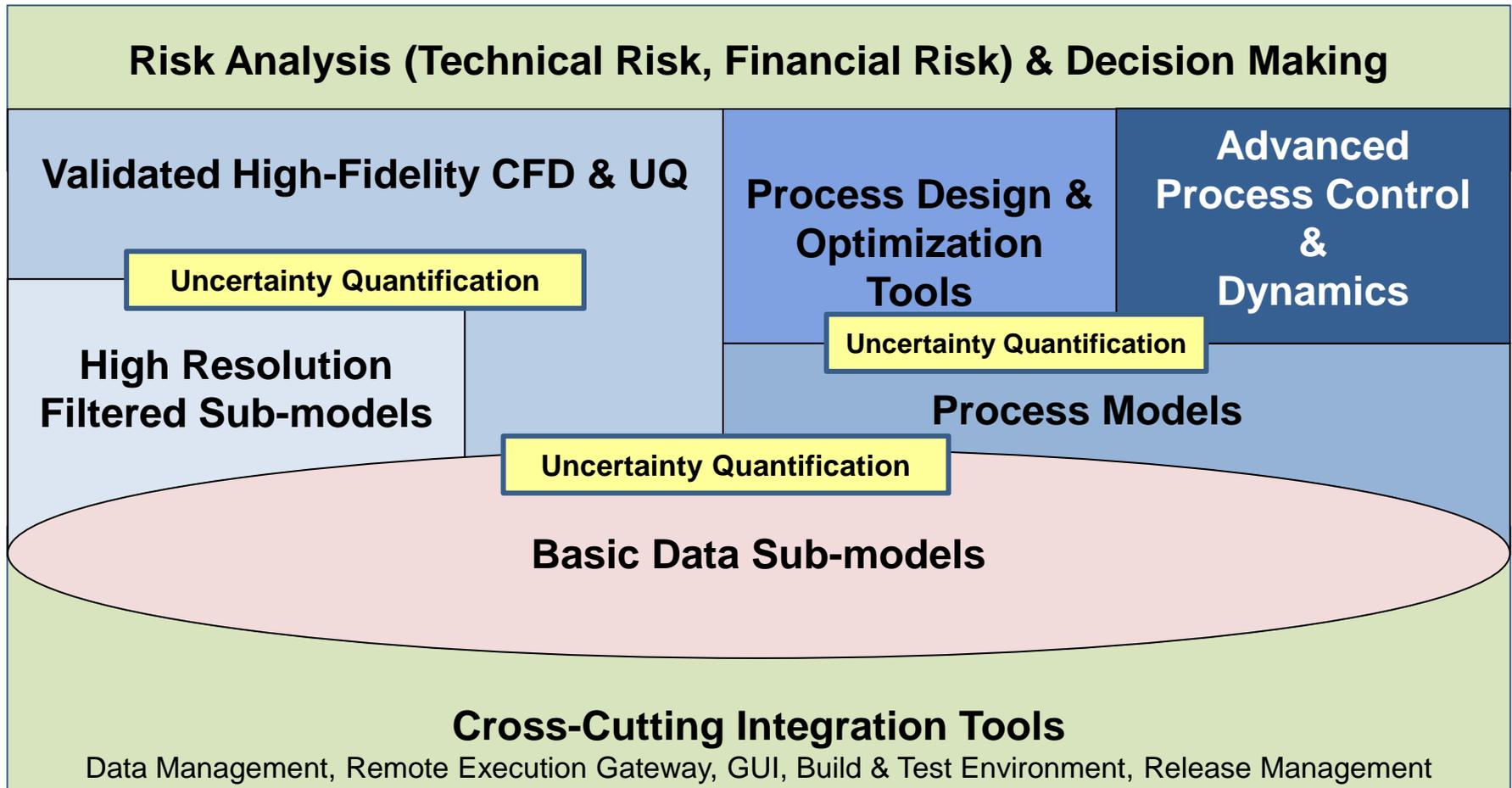
## Academia



## Industry



# Advanced Computational Tools to Accelerate Next Generation Technology Development

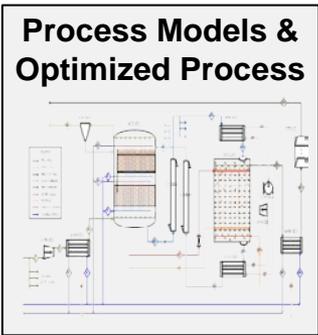


# Simulation & Experiments to reduce time for design/troubleshooting

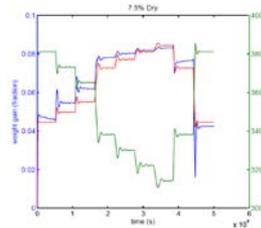
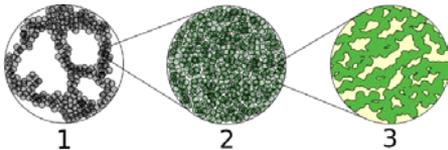
Experimental Validation



Process Models & Optimized Process



**SORBENTFIT**

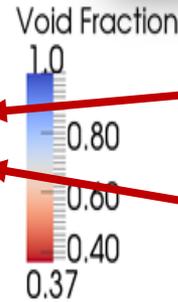


Experimental Kinetic/Mass Transfer Data

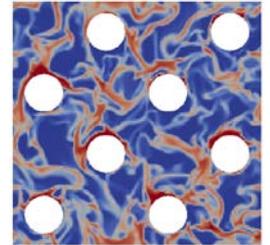
Void Fraction along vertical center plane

**ANSYS**  
FLUENT®

Heat-transfer-tube-scale hydrodynamics

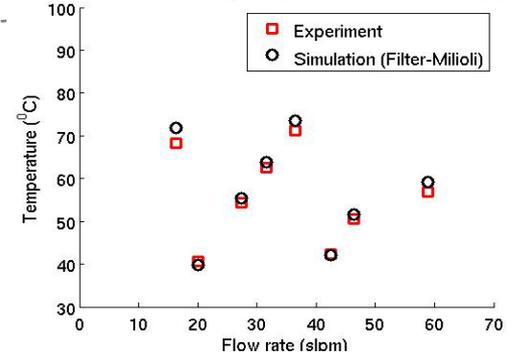


$$f_{drag}^* = \beta^* \left( -v_s^* |v_s^*| \right) + \gamma^*$$



**MFiX**

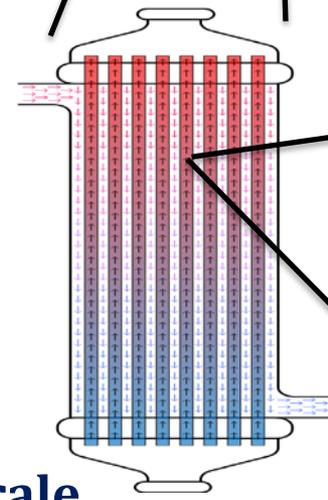
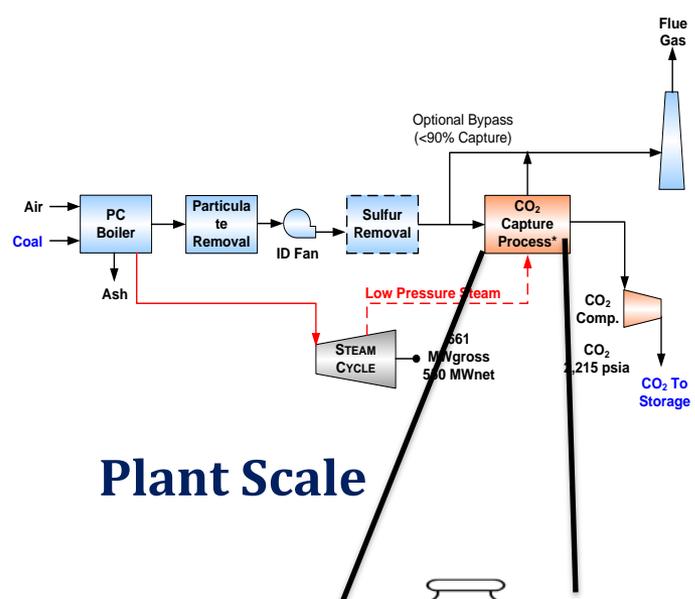
32D Hot Non-reacting Flow (TE3962 & TE3965 average)



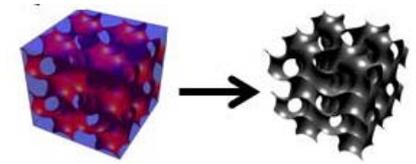
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- **Experimental Carbon Capture Research @ NETL**
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- Conclusions

# Carbon Capture



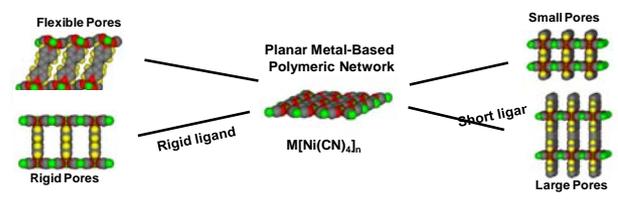
Development of efficient and economic carbon capture strategies applicable for post-, pre- and oxy-combustion schemes



**Polymers and Derivatives**

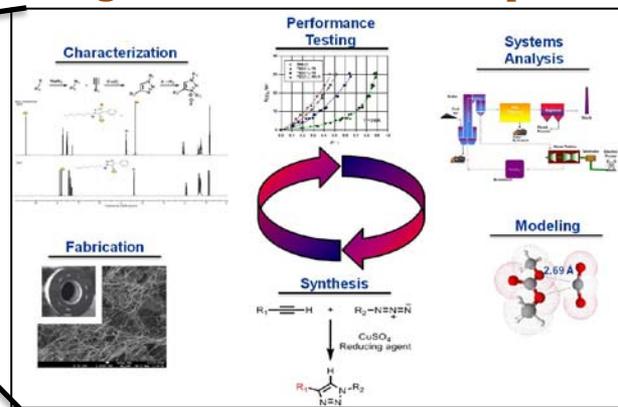


**Ionic Liquids**

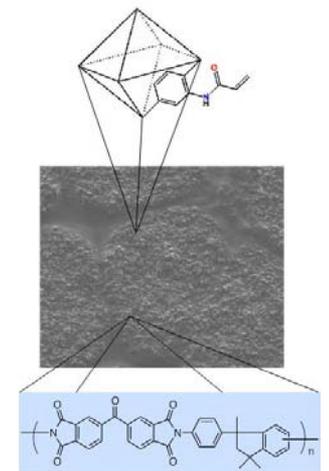


**Metal-Organic Frameworks**

**Integrated Materials Development**



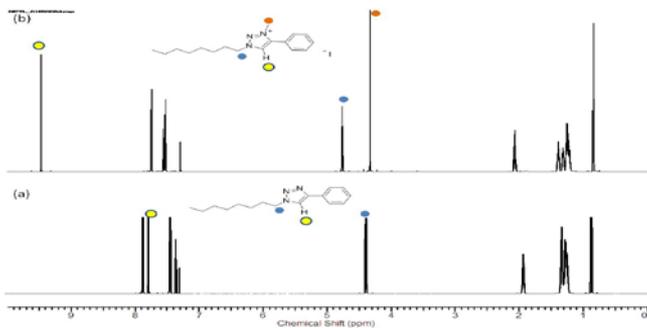
**Materials Scale**



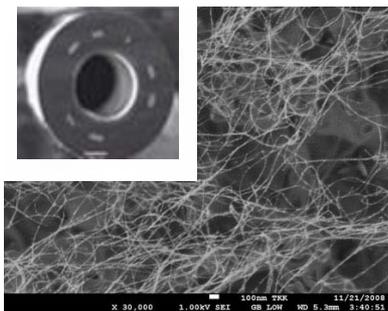
**Composites**

# Integrated Materials Development

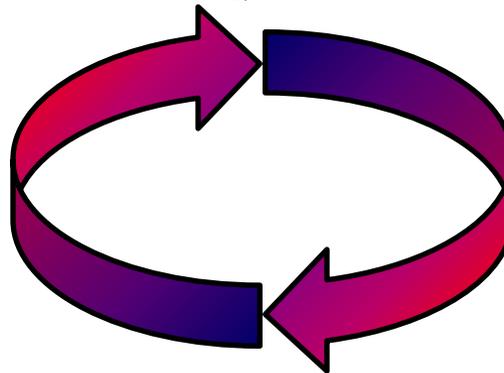
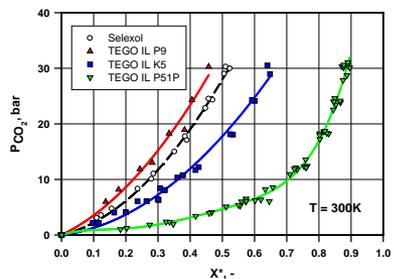
## Characterization



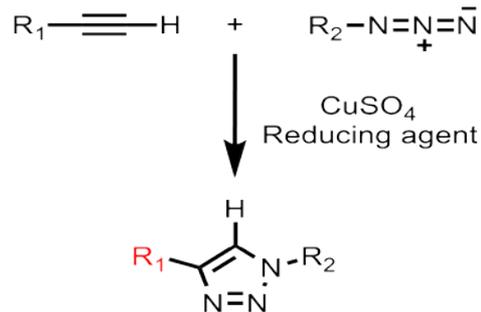
## Fabrication



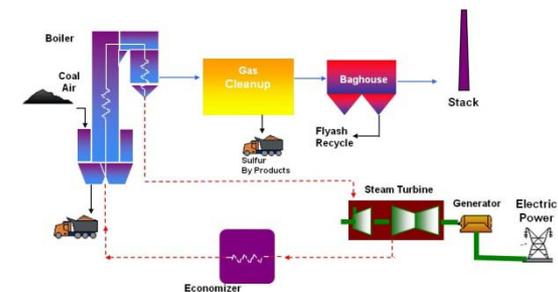
## Performance Testing



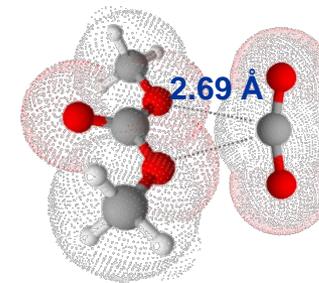
## Synthesis



## CCSI Simulation & Analysis

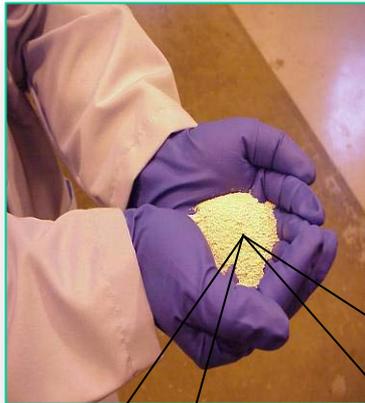


## Molecular Modeling

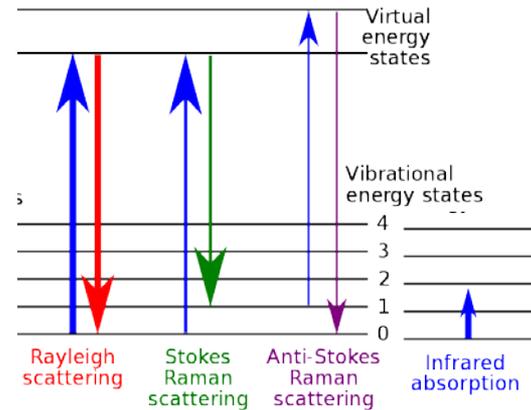


# Carbon Capture

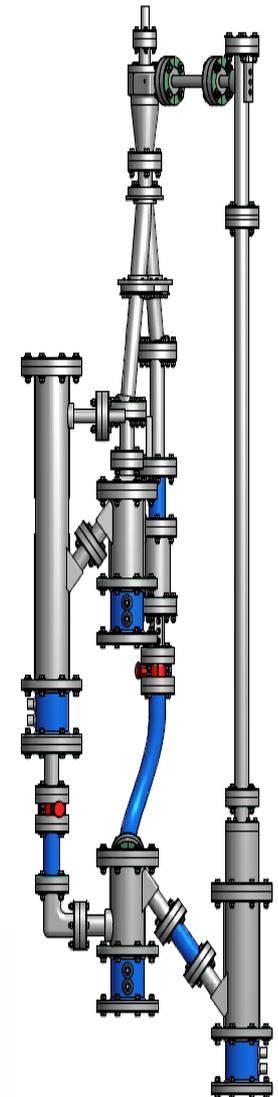
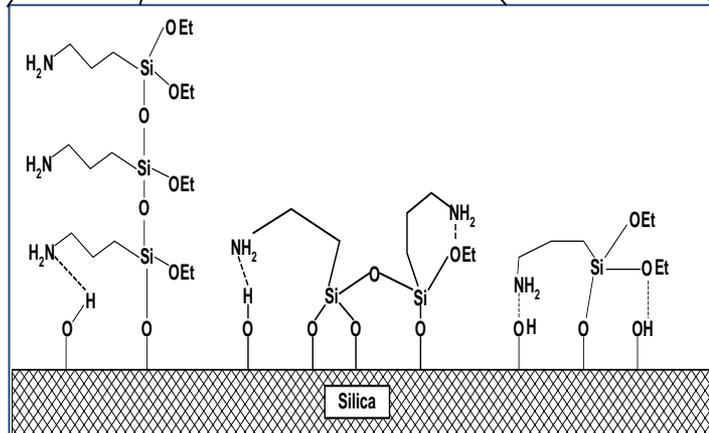
## Supported Amine Sorbents



Characterization

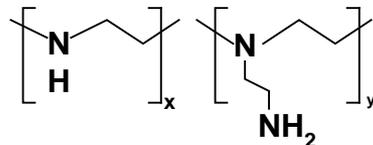


Scale-up

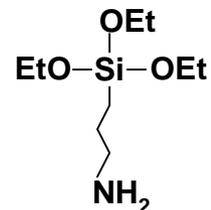


# Components of Basic Immobilized Amine Sorbents (BIAS)

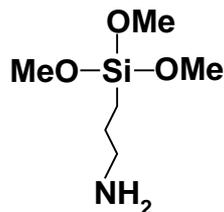
PQ 2129



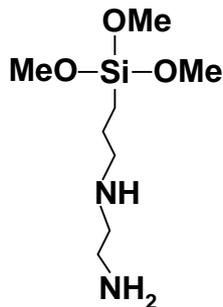
Polyethyleneimine  
PEI  
Mn 423-2000



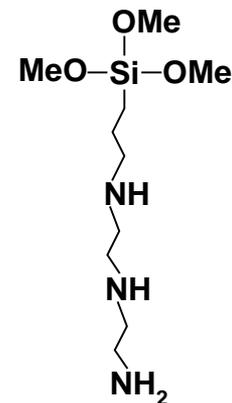
(3-Aminopropyl) triethoxysilane  
APTES



(3-Aminopropyl) trimethoxysilane  
APTMS



N1-(3-Trimethoxysilypropyl)  
diethylenediamine  
DEDA



N1-(3-Trimethoxysilypropyl)  
diethylenetraamine  
DETA

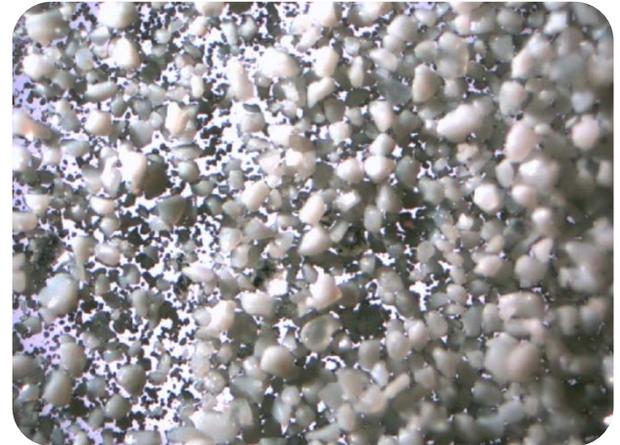
Modification of US 7,288,136 High Capacity  
Immobilized Amine Sorbents  
US Patent Application 13212284 filed 8/11

# Sorbent AX

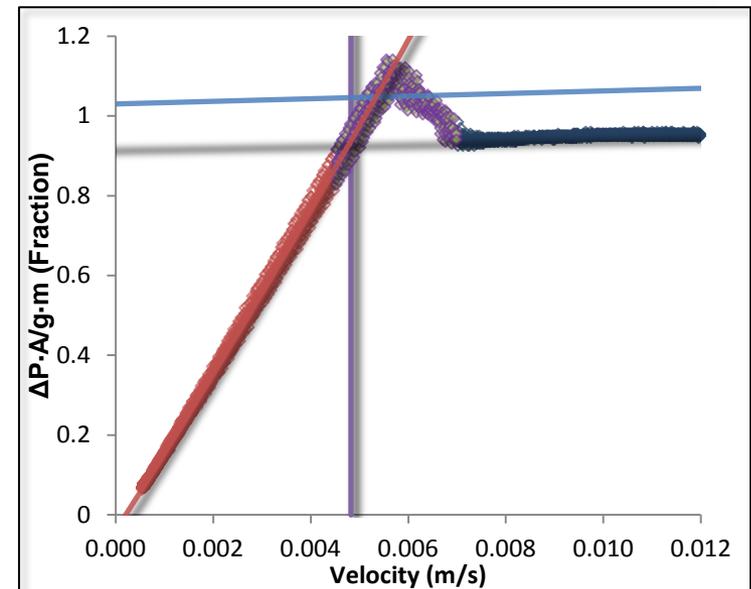
Two BIAS sorbents are being tested: AX and 32D

## Sorbent AX

- 40% PEI - BASF M<sub>n</sub> 2000
- Mesoporous Silica support- PQ Inc 2129
- 8 - 15 gal. drums from ADA
- Capacity (avg.) = 2.82 mmole CO<sub>2</sub>/g adsorbent ranging from 2.60-2.87 for n=7
- ADA packed bed, 13-14% CO<sub>2</sub>, ~55°C Capacity ( avg.) =1.25 avg. mmole CO<sub>2</sub>/g adsorbent
- $u_{mf}=0.48$  cm/s
- Particles behavior is Geldart Group A

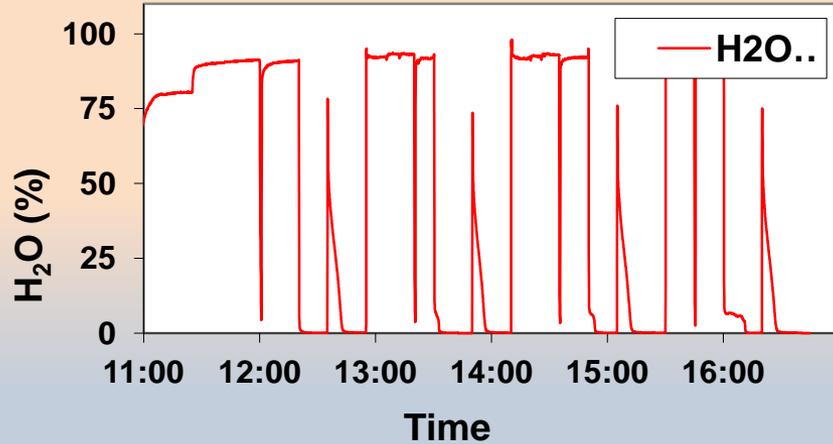


Sauter mean particle diameter	( $\mu\text{m}$ )	114
Spericity	(UNITLESS)	0.86
Particle porosity	(UNITLESS)	0.39
Particle skeletal density	(g/cc)	1.50
Particle density	(g/cc)	0.91

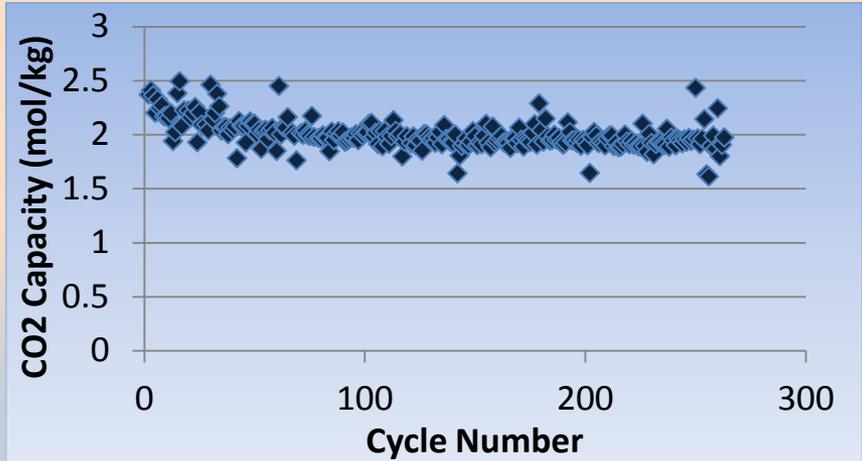


# BIAS Sorbent Testing

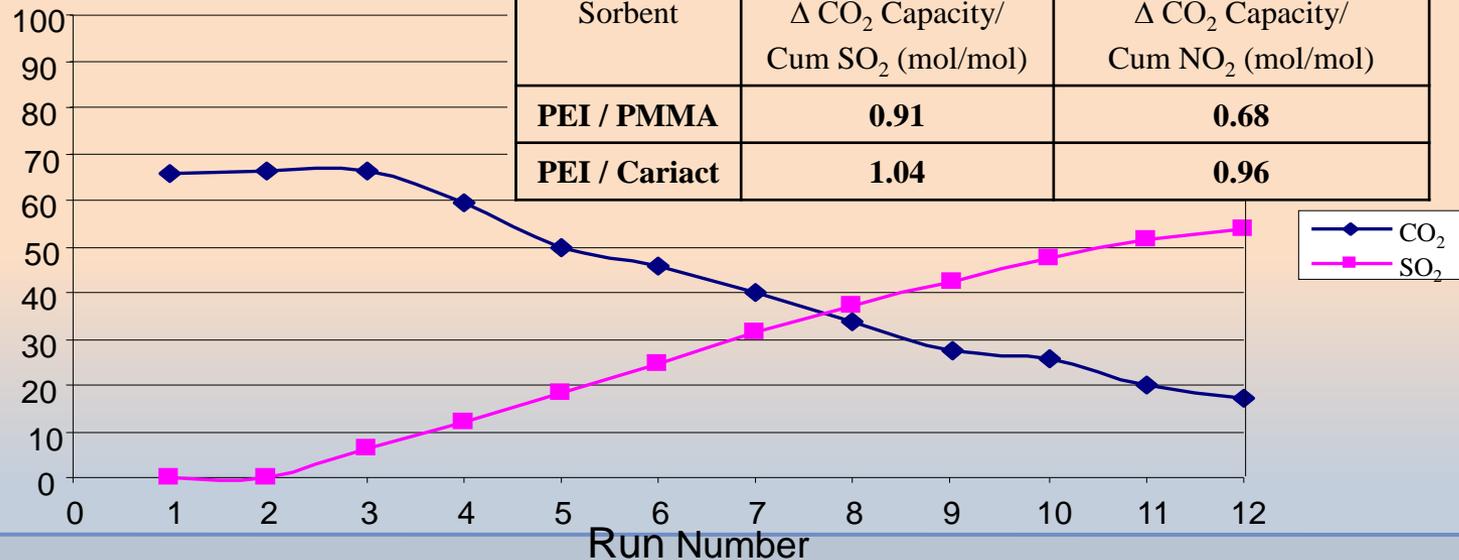
Regeneration at 105 °C with 90% steam



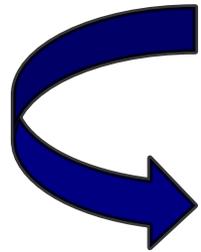
250 cycle stability under humidified conditions



Accelerated SO<sub>x</sub> testing



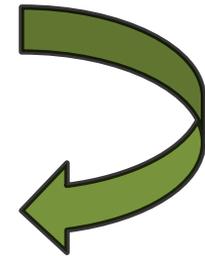
# CCSI communication



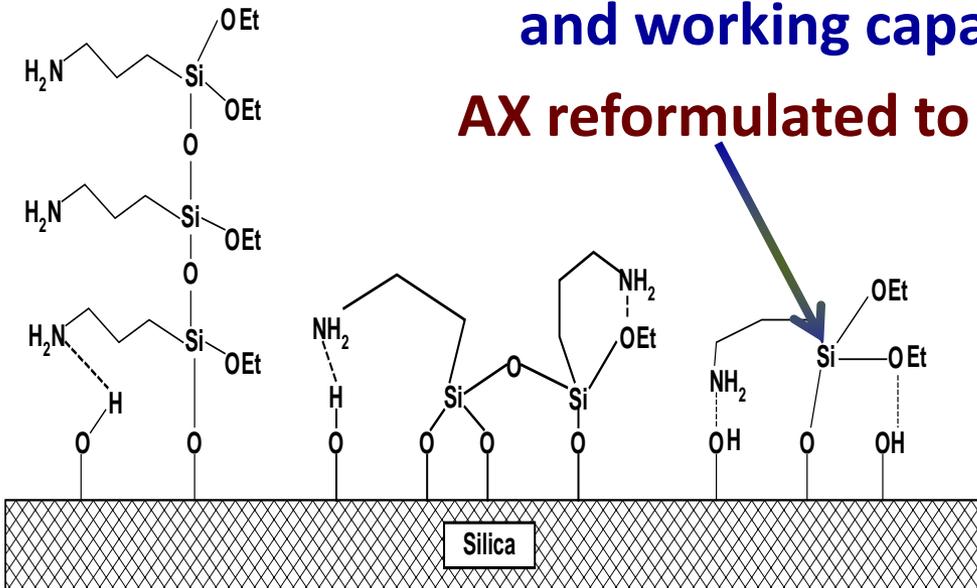
**Provided AX sorbent properties to CCSI**

**Feedback on working capacity and moisture requirements**

**Reformulates sorbent based moisture and working capacity requirements**

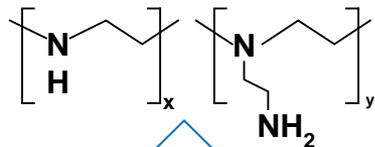


**AX reformulated to 32D sorbent**



# Polyethyleneimine Silane Coupling

Polyethyleneimine Mn 423-2000



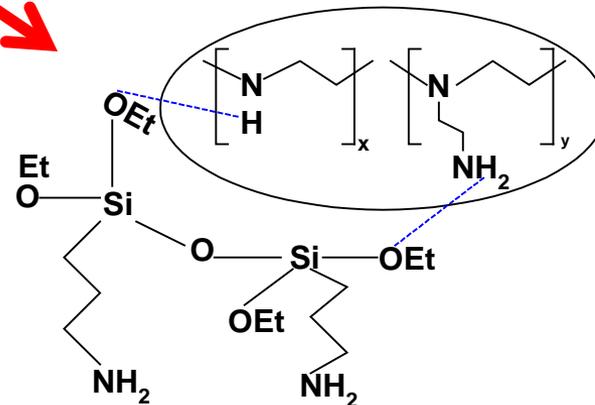
Aminosilanes

**Simple**  
**Scalable**  
**Acceptable Capacity**  
**Moisture Resistance**  
**Stability**  
**Saleable**

Methanol 80 C  
Vacuum atm- 10 mm Hg

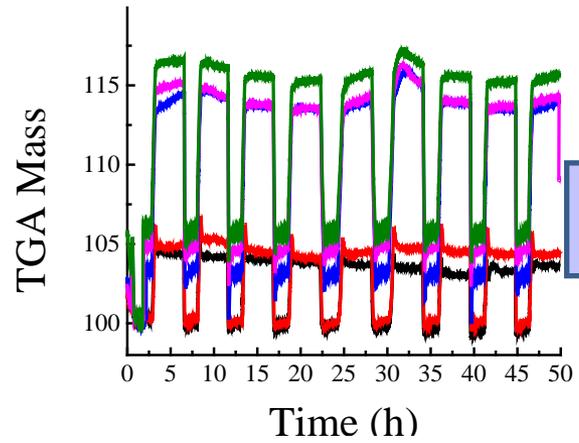
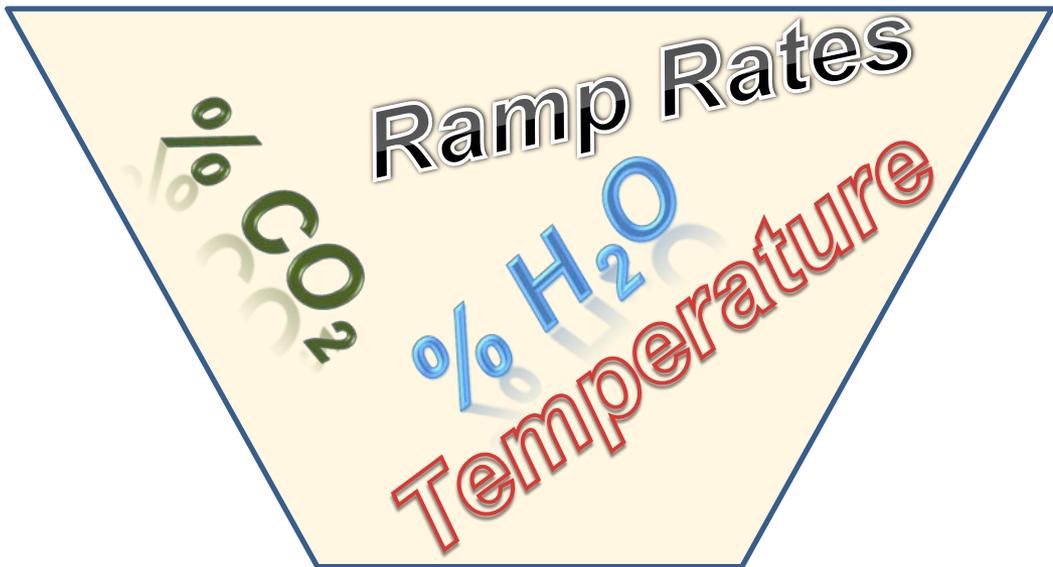


Pressure Chemical – Pan Dyer



Synthesis was scaled to 1,000 lb range

# TGA Data Collection



*Sorption Kinetics*

# Summary for Basic Immobilized Amine Sorbent

- High delta loadings in the 3-4mol/kg range
- CO<sub>2</sub> regeneration improbable
- Stable at elevated temperatures
- Silica substrate candidate of choice
- Loading results confirmed by TVA and ADA-ES
- Moisture adsorption may impact energetics
- Susceptible to poisoning with SO<sub>2</sub> and NO<sub>2</sub>; upstream cleaning required in process
- Kinetic study conducted



PEI on CARIACT Q10  
(100 to 350  $\mu\text{m}$  dia.)



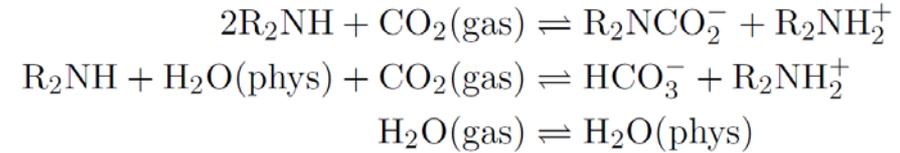
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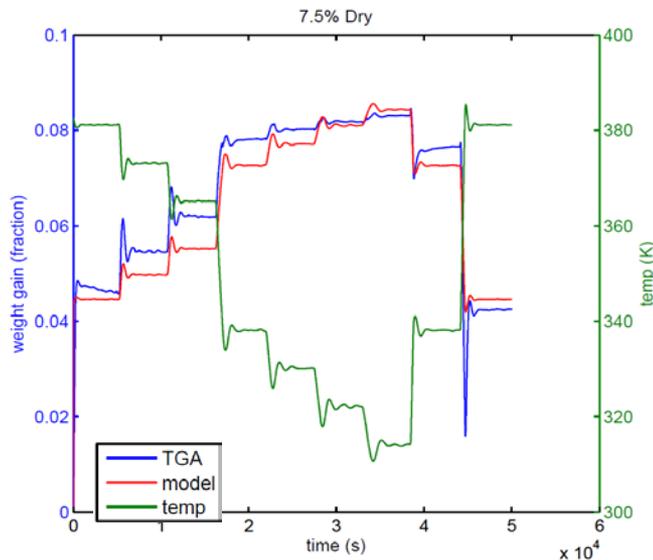
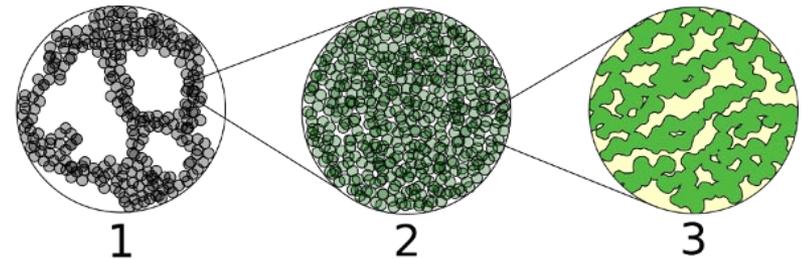
# PEI-Impregnated Silica Sorbent Reaction Model

- **A general lumped kinetic model, quantitatively fit to TGA data, needed for initial CFD and process simulations**



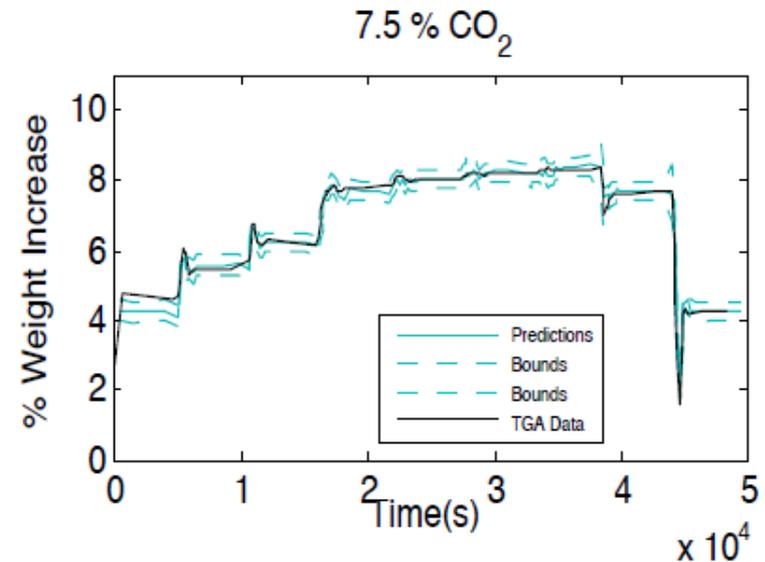
- **High-fidelity model:**

- Sorbent microstructure broken down into three length scales
- Rate of reaction controlled by the diffusion of  $CO_2$  through the amine polymer
- Ab initio calculations indicate dependence of the diffusion process on water



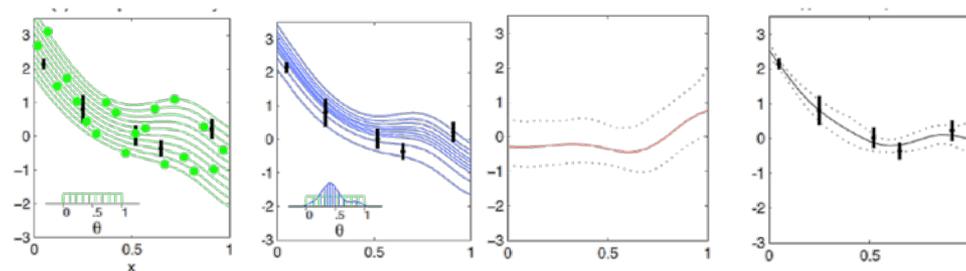
*(left) lumped kinetic fit to experimental TGA for NETL-32D sorbent*

*(right) calibrated model with discrepancy and error bounds*

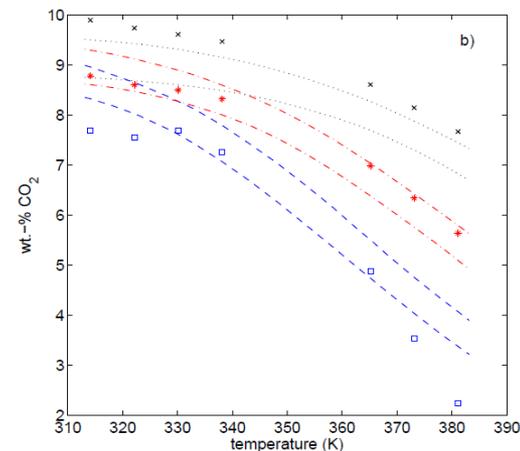
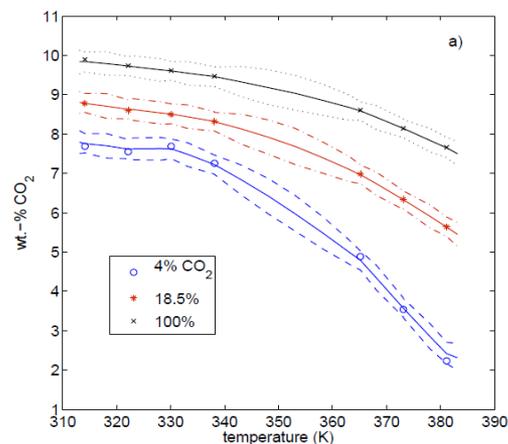


# Bayesian Methods in Parameter Estimation

- *Experimental data tends to constrict the prior distribution, resulting in an experiment-based estimate influenced by theoretical calculations.*
- *A stochastic function representing the model error can also be estimated in this way.*



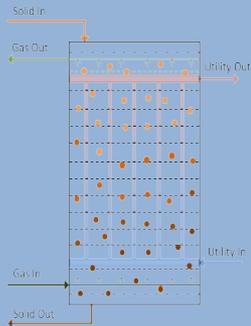
Above: schematic of the calibration process. Left to right: draws from the prior, draws from the posterior, discrepancy, and predictions.



Right: model-plus-discrepancy (a) and model-only predictions (b), with confidence bounds.

# CCSI Tools to develop an optimized process using rigorous models

## Process Models



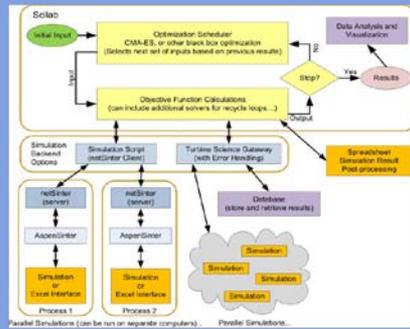
## Basic Data Submodels



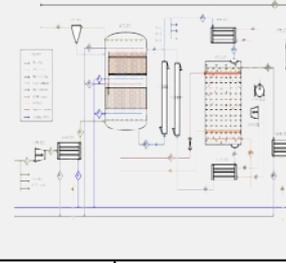
## Algebraic Surrogate Models



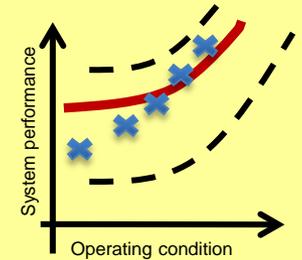
## Simulation-Based Optimization



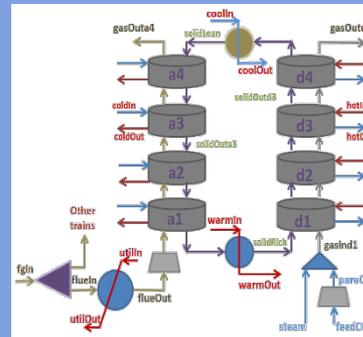
## Optimized Process



## Uncertainty Quantification



## Superstructure Optimization (Determine Configuration)



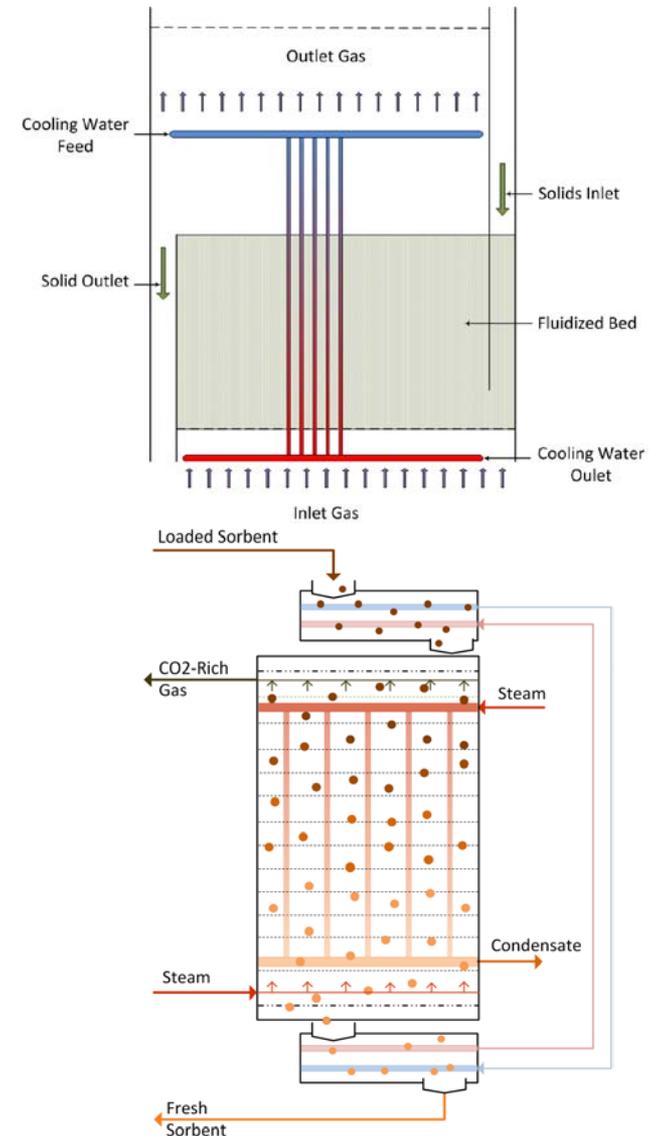
# Solid Sorbent System Models

## Bubbling Fluidized Bed (BFB) Models

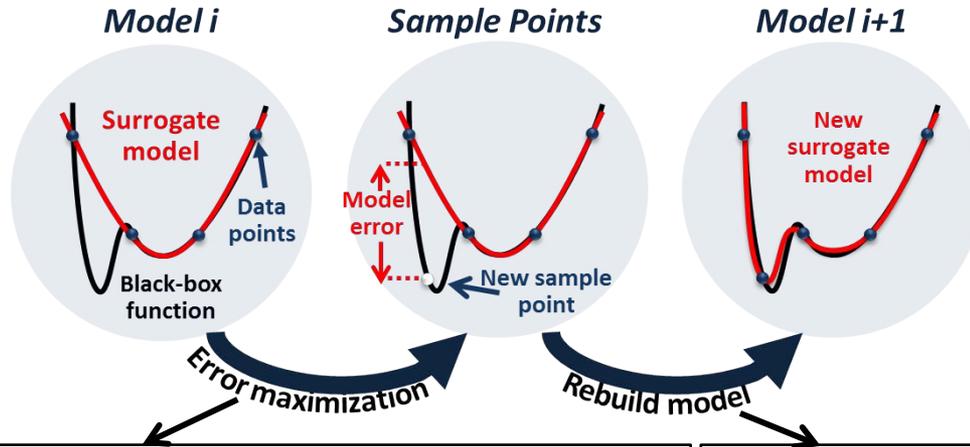
- Flexible BFB models with immersed heat exchangers have been developed to be used as adsorber or regenerator, as needed, with varying locations for solids inlet and outlet streams
- Any number of BFB adsorbers and/or regenerators can be connected in series and/or in parallel depending on the user requirements
- A 2-stage adsorption model with customized variables suitable for incorporating UQ has been developed

## Moving Bed (MB) Models

- External resistance to mass transfer has been modeled. This is particularly important in the regenerator model due to the high operating temperature.
- Heat exchanger model, mass and heat transfer coefficients, boundary conditions, temperature specifications, and properties models are revisited for better model accuracy.



# Automated Learning of Algebraic Models for Optimization



**For building accurate, simple algebraic surrogate models of simulated processes**

$$\max_x \left( \frac{z(x) - \hat{z}(x)}{z(x)} \right)^2$$

Labels:  $\hat{z}(x)$  is the **Surrogate model**,  $z(x)$  is the **Simulation/black-box**.

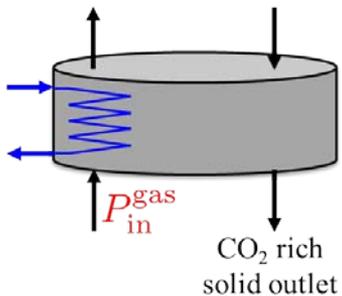
**Step 1: Define a large set of potential basis functions**

$$\hat{z}(x) = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_1 x_2 + \beta_4 \frac{x_1}{x_2} + \beta_5 \frac{x_2}{x_1} + \beta_6 e^{x_1} + \beta_7 e^{x_2} + \dots$$

**Step 2: Model reduction**

$$\hat{z}(x) = \beta_0 + \beta_2 x_2 + \beta_5 \frac{x_2}{x_1} + \beta_7 e^{x_2}$$

## Example Model: BFB Adsorber Inlet Gas Pressure



- **ACM Simulation**
- **>900 terms possible**
- **14 input variables**
- **0.13% error**

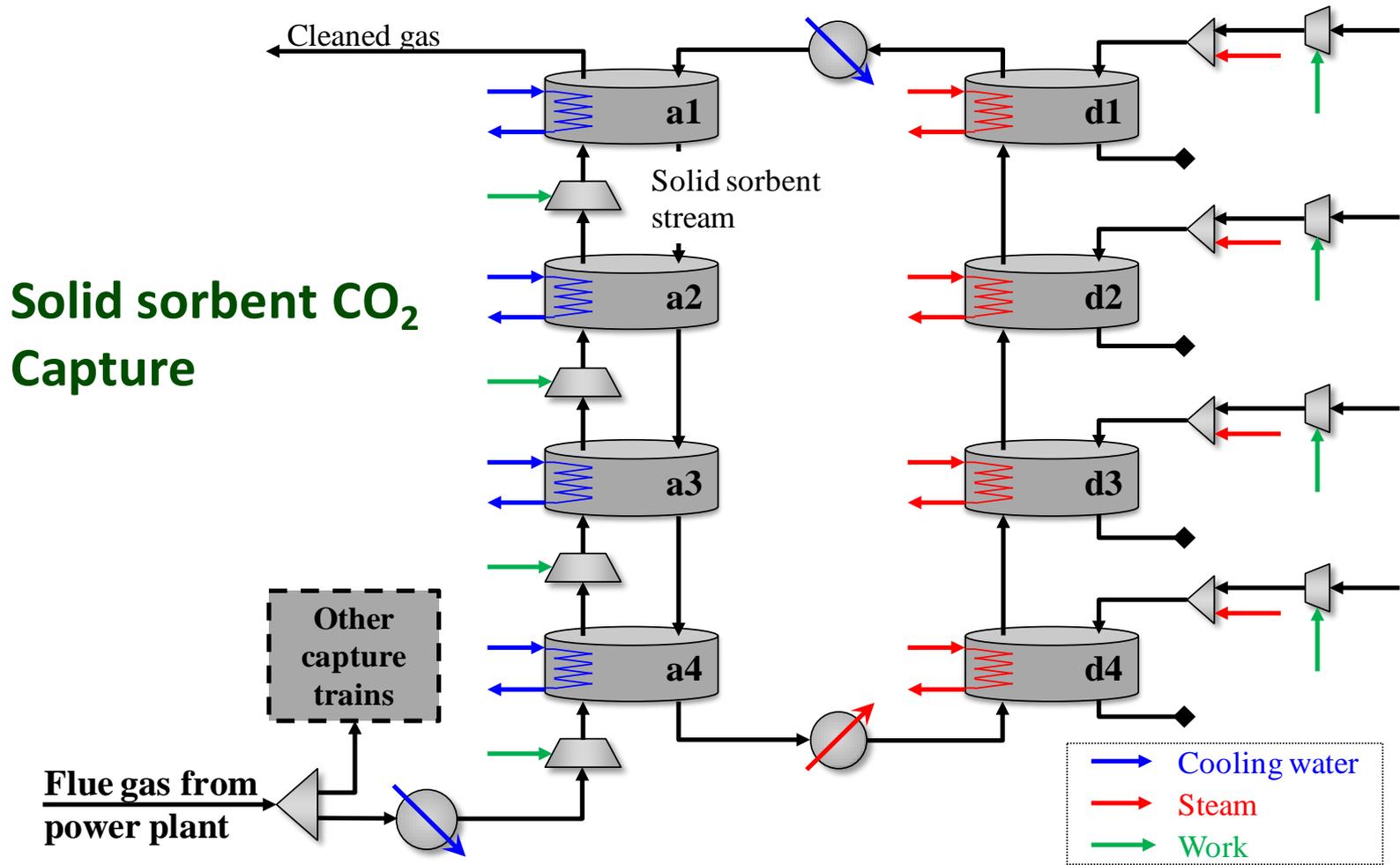
Pressure drop across length of bed

$$\hat{P}_{in}^{gas} = P_{out}^{gas} + 0.019 L_b + 0.0055 \sqrt{D_T}$$

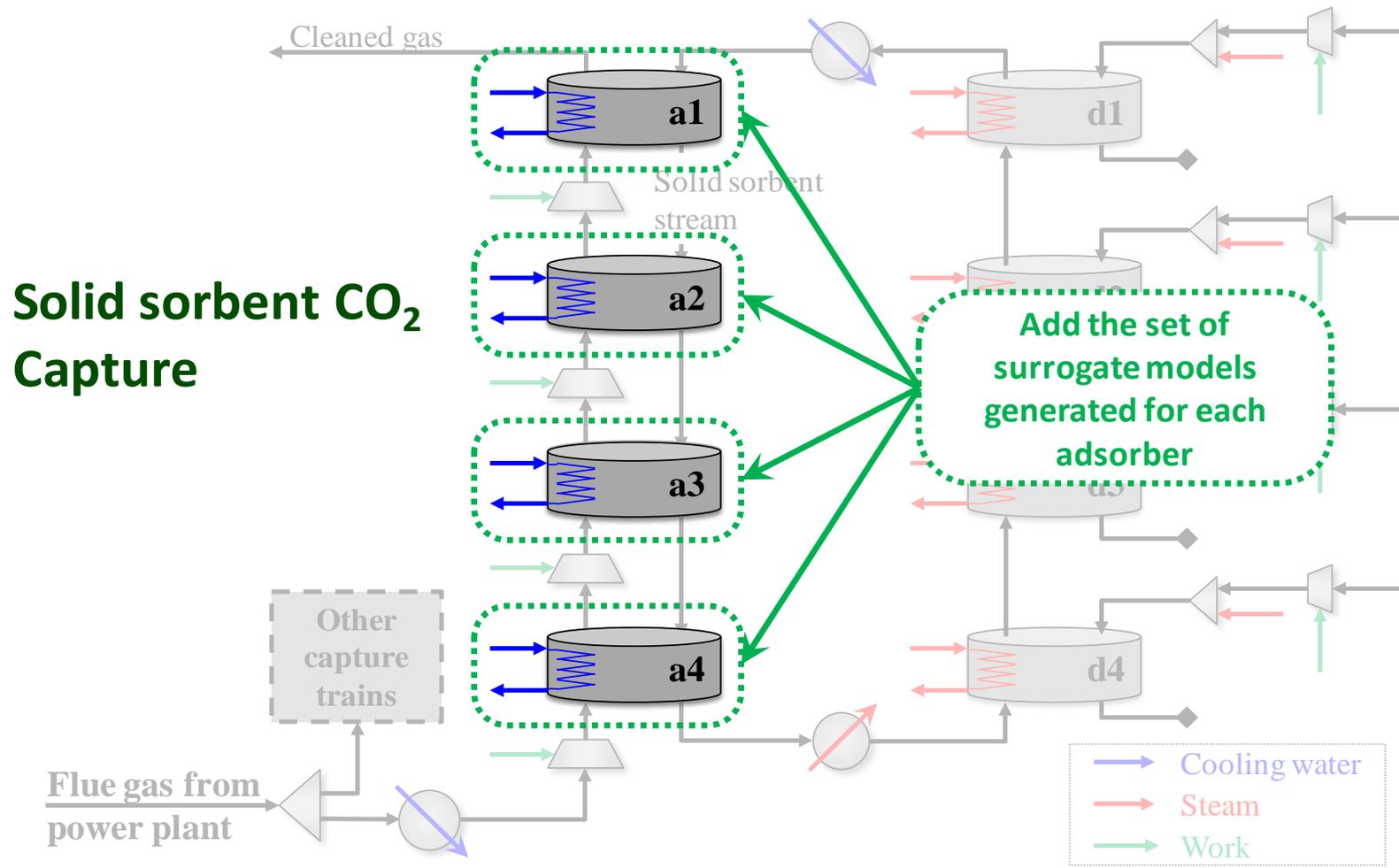
Proportional to outlet pressure

Pressure drop due to bed diameter

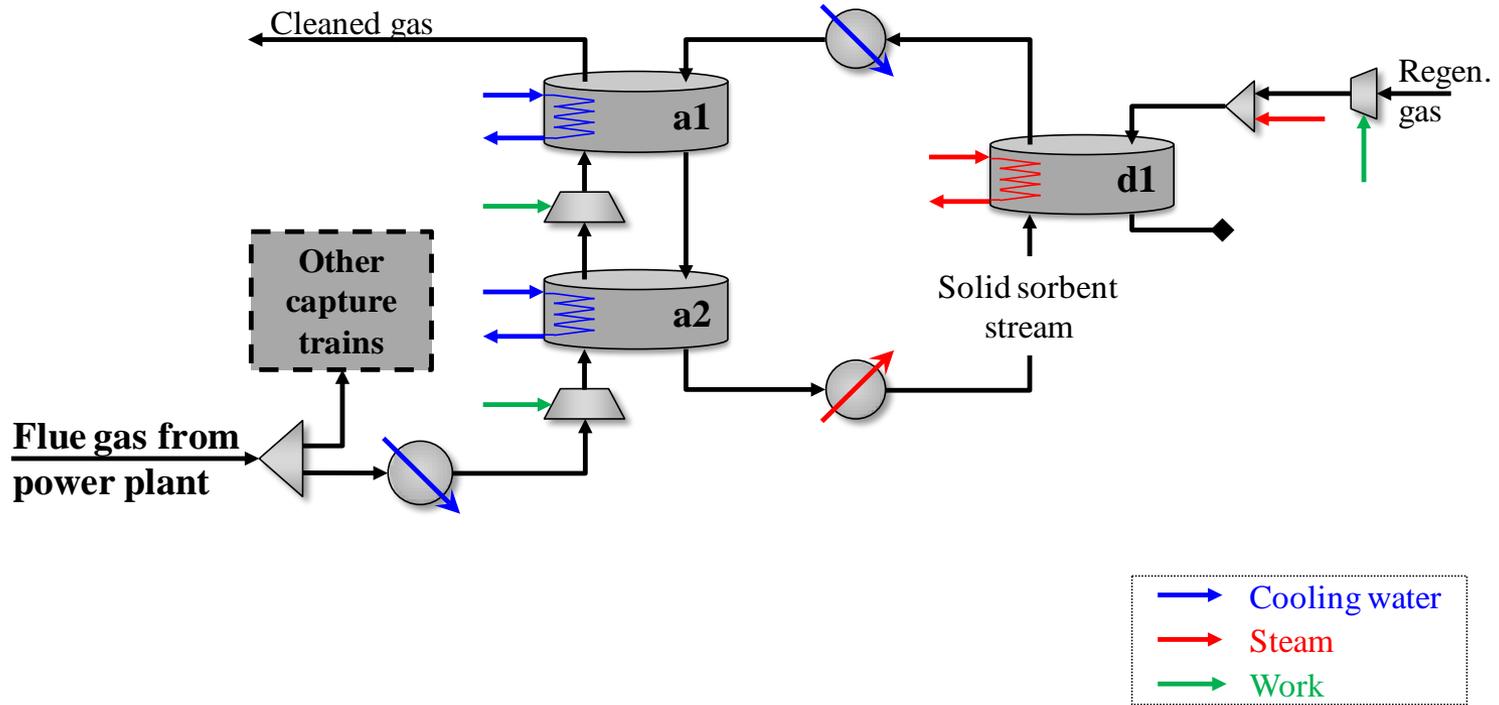
# Superstructure Formulation & Optimization



# Insert Algebraic Surrogates into Superstructure

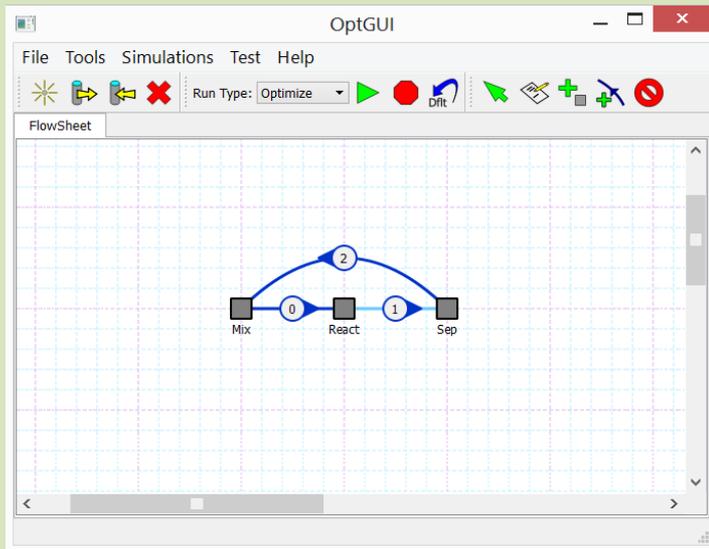


# Initial Superstructure Solution



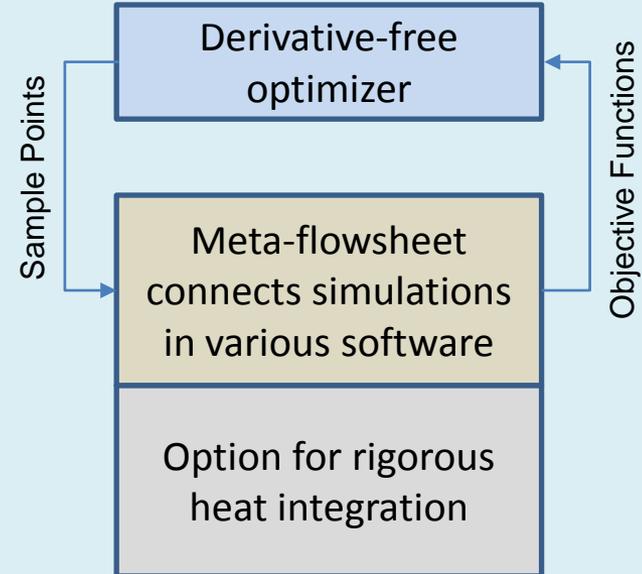
# Simulation-Based Optimization Framework

## Graphical Interface



**Turbine Gateway**  
(runs simulations and stores results  
can run simulations in parallel)

## Optimization Engine



# Turbine Science Gateway

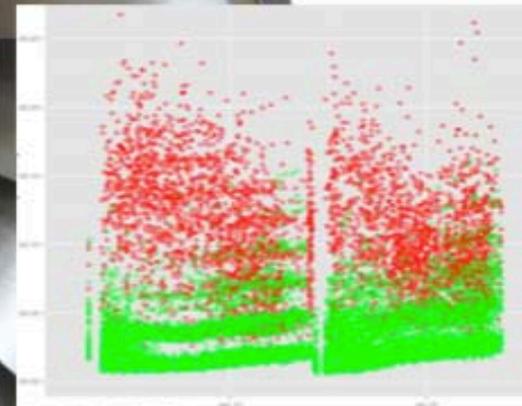
## Turbine Science Gateway

Parallel Simulation Execution  
SaaS On-Demand Provisioning  
Amazon EC2 Cloud

### ACM Hybrid Split Optimization

- Experiment ran 13000 simulations
- 100 simulations per iteration
- 130 iterations Over 2 days using 50 virtual machines

PSE gPROMS is now supported. Deployed on EC2 with 5 gORun\_xml licenses



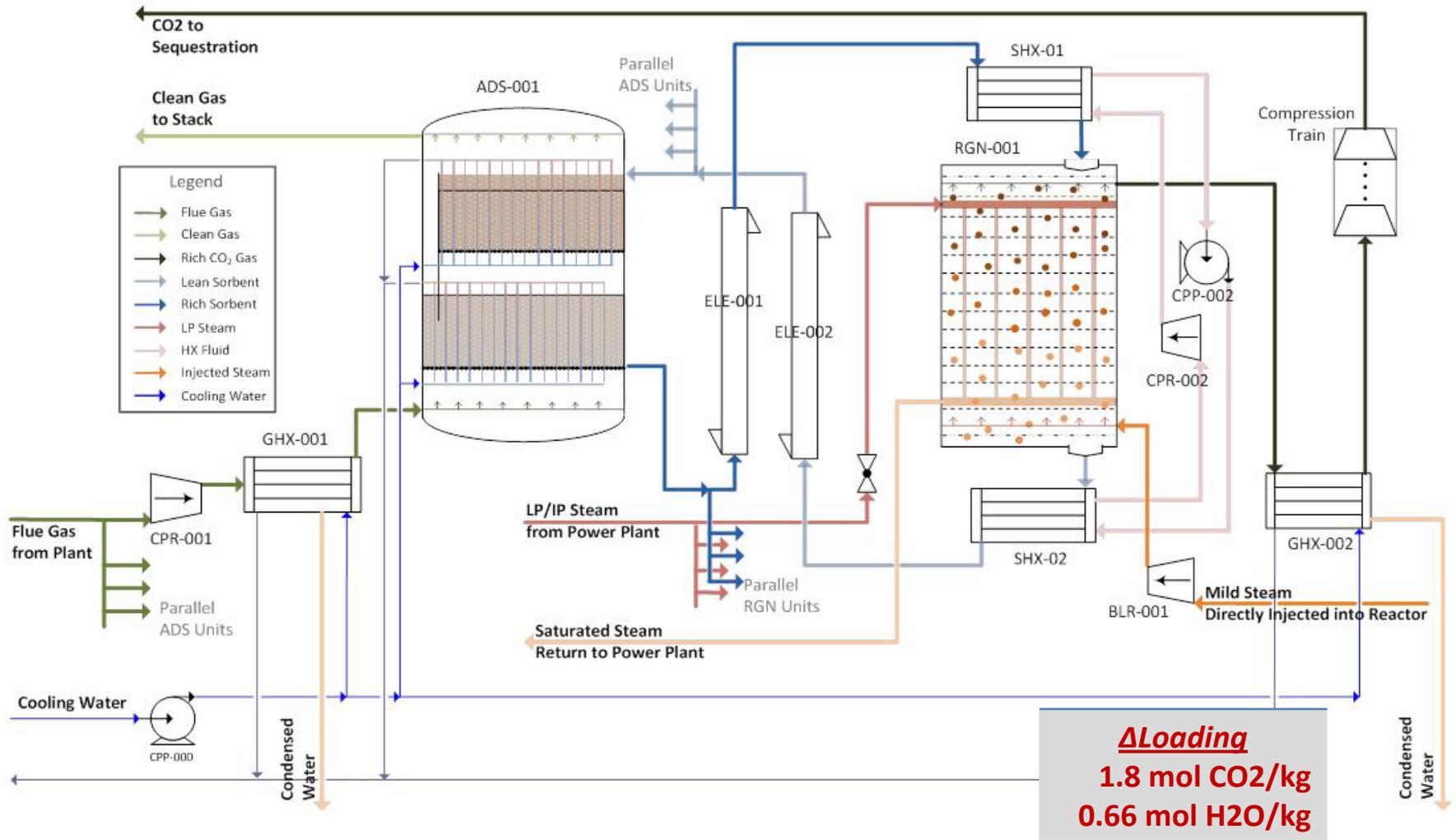
Data Management  
Real-time Log  
analysis of parallel  
simulations



# Decision Variables

<b>Input Variable</b>	<b>Lower Bound</b>	<b>Upper Bound</b>
Adsorber Diameter (m)	7	10
Top & Bottom Adsorber Bed Depth (m)	4	10
Top & Bottom Adsorber Heat Exchanger Tube Diameter (m)	0.01	0.05
Top & Bottom Adsorber Heat Exchanger Tube Pitch (m)	0.1	0.2
Top & Bottom Adsorber Cooling Water Flowrate (kmol/hr)	30,000	60,000
Sorbent Flowrate per Adsorber (kg/hr)	350,000	600,000
Gas Pre-Cooler Temperature (°C)	40	60
Regenerator Height (m)	3	7
Regenerator Diameter (m)	6	10
Regenerator Heat Exchanger Tube Diameter (m)	0.01	0.05
Regenerator Direct Steam Injection Rate (kmol/hr)	900	1400
Regenerator Heat Exchanger Steam Flowrate (kmol/hr)	2,500	5,000

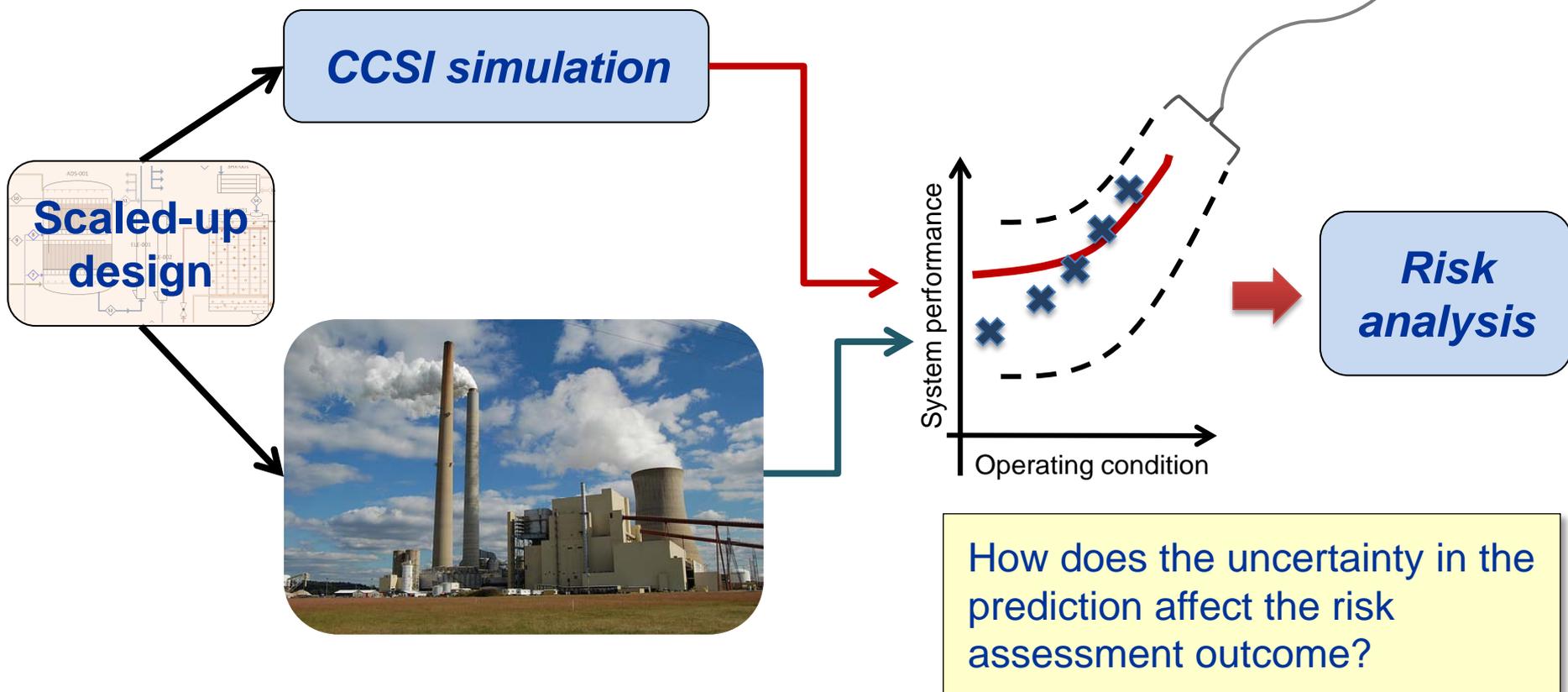
# Optimized Capture Process Developed using CCSI Toolset



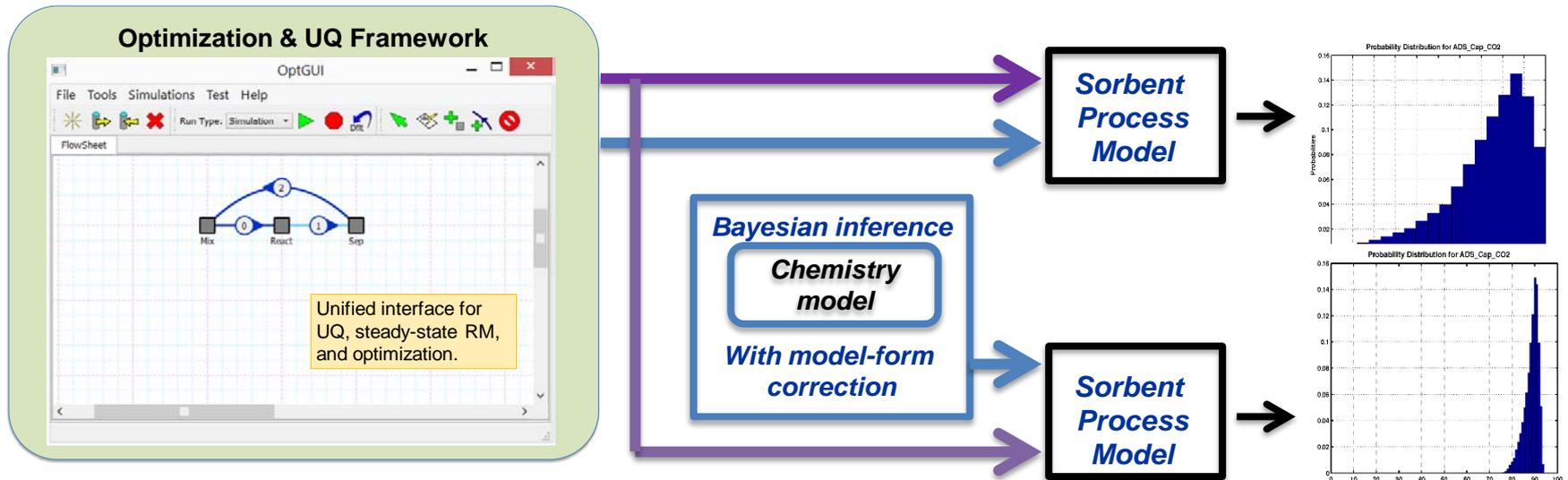


# Uncertainty Quantification: How certain are we that our model can predict the system performance accurately?

- How to quantify these error bounds *a priori*?
- How to reduce these bounds?

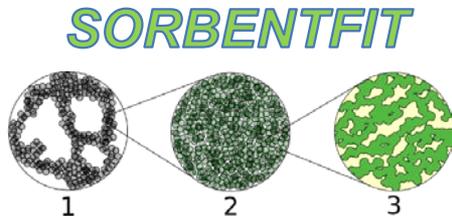
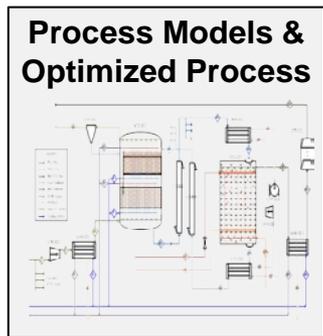


# Multi-Scale Uncertainty Quantification Framework

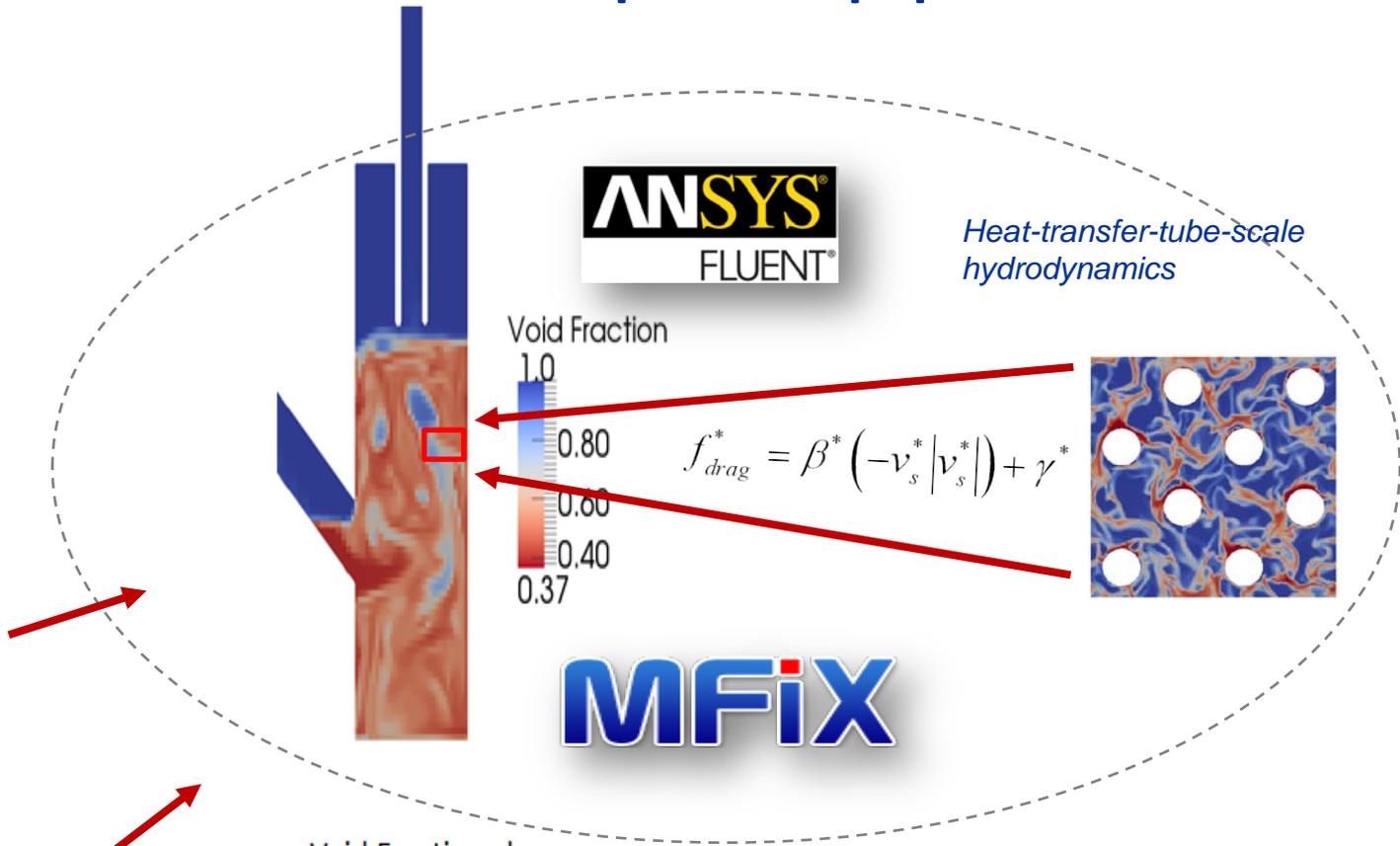
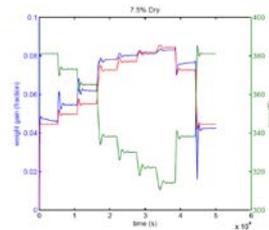


- **UQ for basic data models**
  - Bayesian UQ methodology
  - Integration of model form discrepancy into process & CFD models
- **UQ for CFD models**
  - Adaptive sampling capability for RM/UQ
  - Bayesian calibration capability
  - UQ of discrepancy between CFD/process models
- **UQ for process models**
  - Integration with optimization platform
  - Optimization under uncertainty

# Detailed CFD Simulations of Specific Equipment



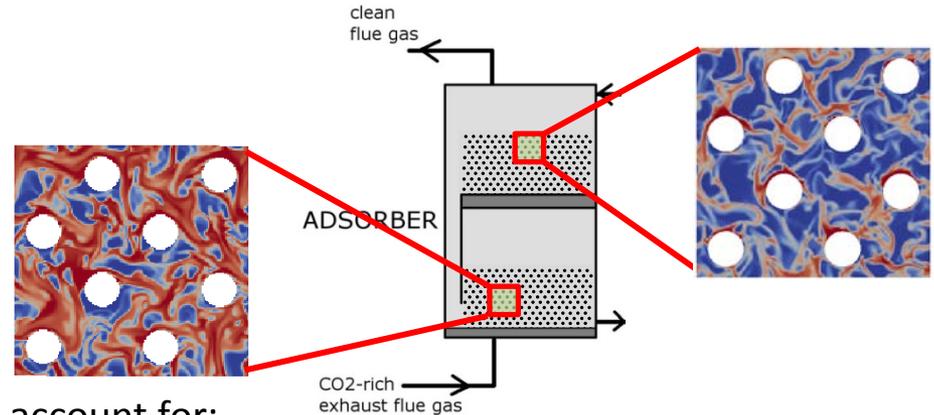
Experimental Kinetic/Mass Transfer Data



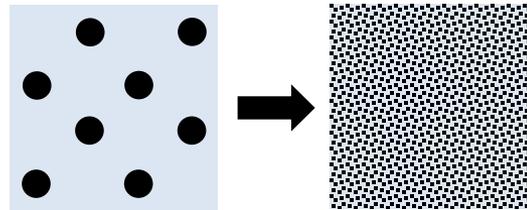
Void Fraction along vertical center plane

# High Resolution Particle Models

- **Problem:** Explicit resolution of small-scale particle clusters and cooling tubes in large devices computationally infeasible!



- **Solution** – Develop sub-grid ‘filtered’ that account for:
  - Unresolved particle clusters.
  - Drag exerted by cylinders on suspension. **Work well along the way for horizontal tubes**

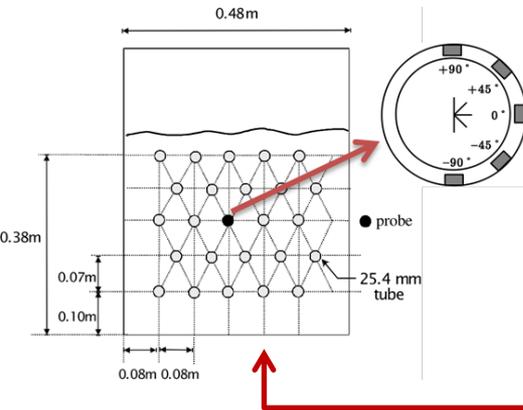
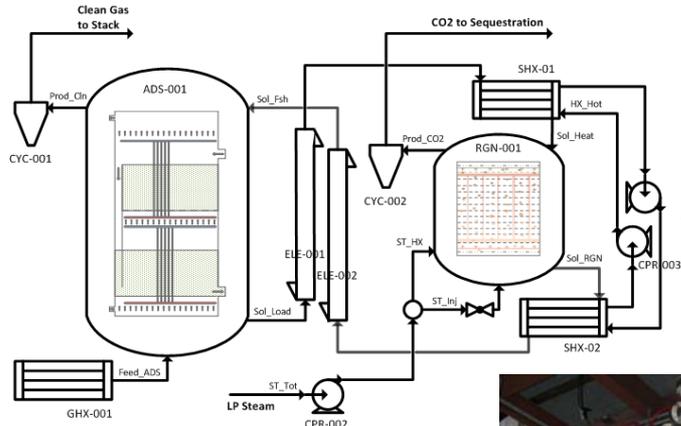


*Cylinders replaced by an **equivalent** stationary, porous medium.*

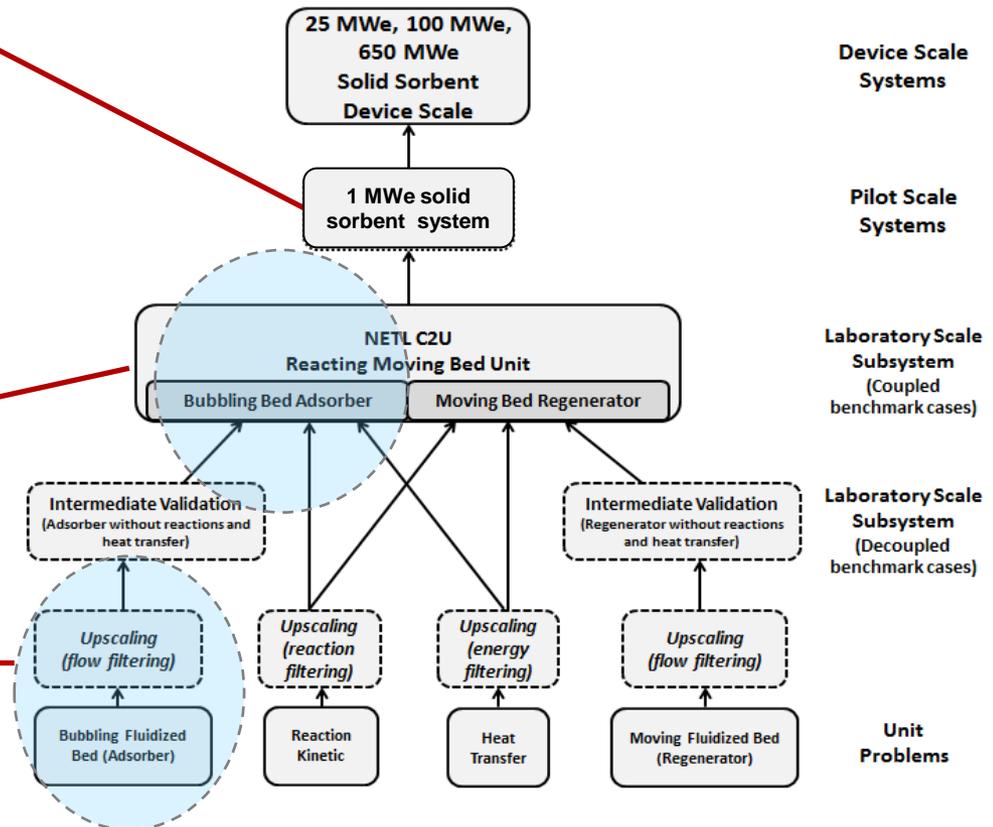
- **Benefit:** The sub-grid ‘filtered’ models can be implemented in faster, coarse-grid simulations.

# Validated CFD Models at the Device Scale

**Objective: To provide quantitative confidence on device-scale (CFD) model predictions for devices that are yet to be built.**



## CCSI CFD Validation Hierarchy

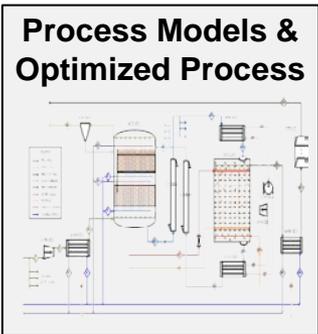


# Rigorous Validation of Models

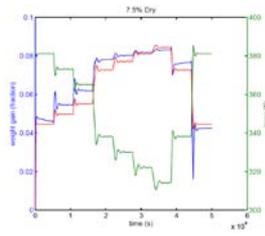
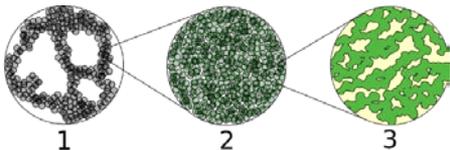
Experimental Validation



Process Models & Optimized Process



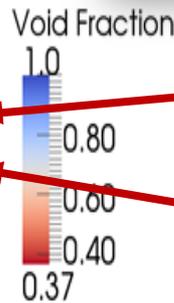
**SORBENTFIT**



Experimental Kinetic/Mass Transfer Data

Void Fraction along vertical center plane

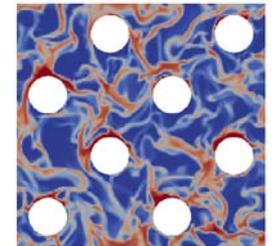
**MFiX**



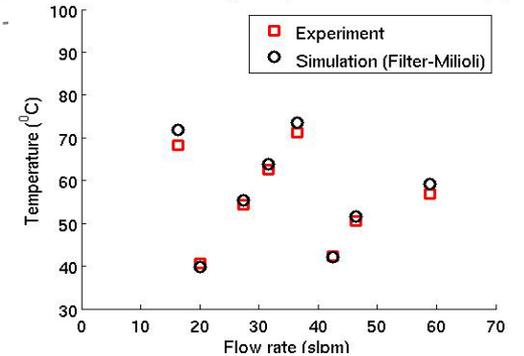
**ANSYS**  
FLUENT®

Heat-transfer-tube-scale hydrodynamics

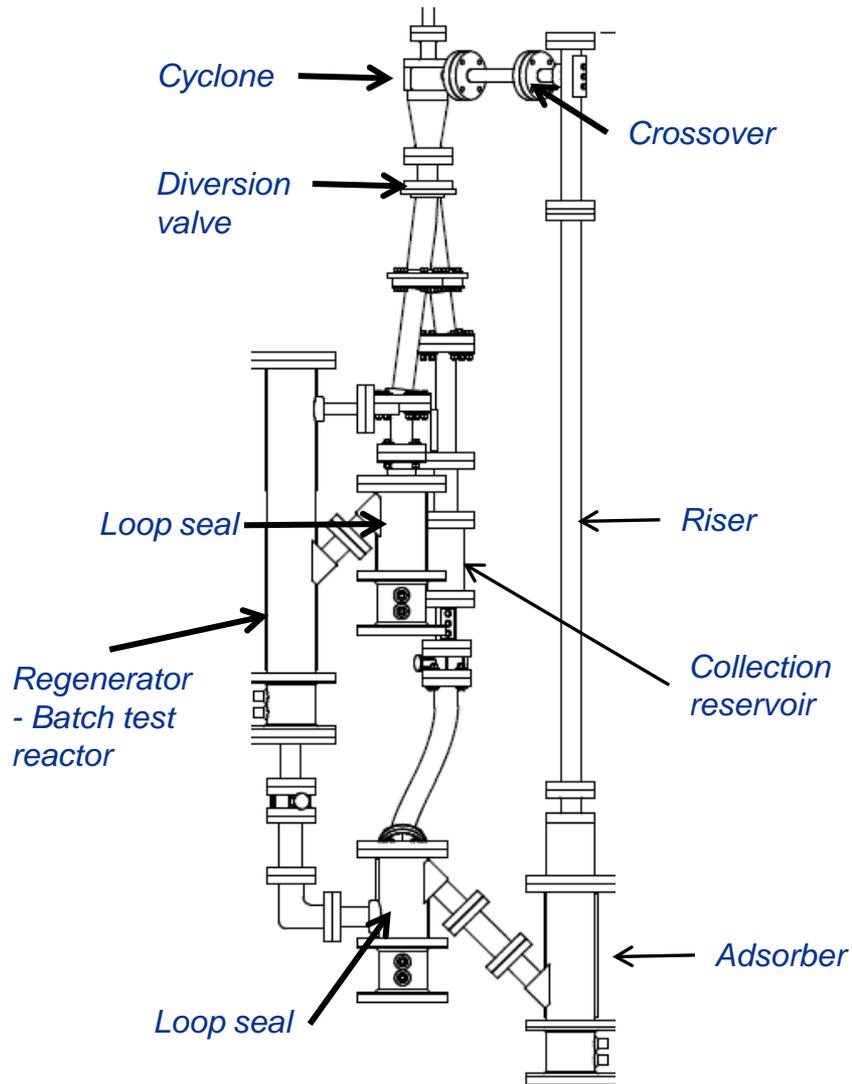
$$f_{drag}^* = \beta^* \left( -v_s^* |v_s^*| \right) + \gamma^*$$



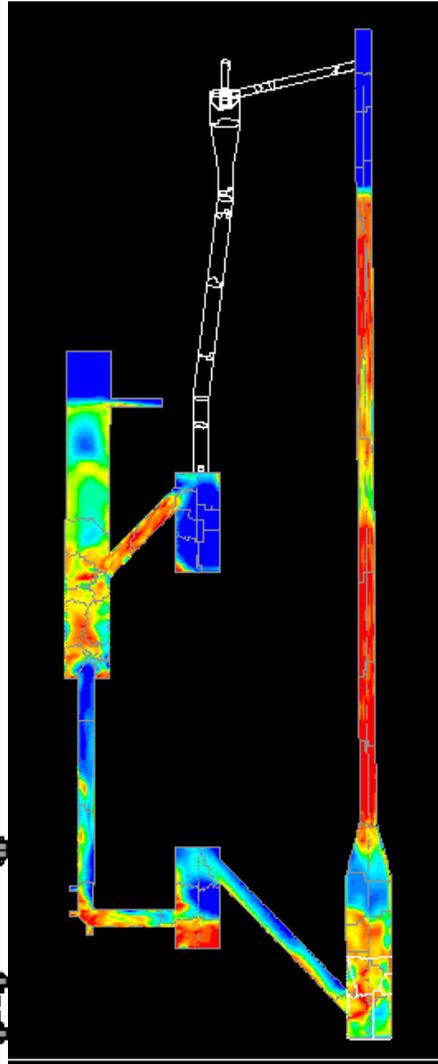
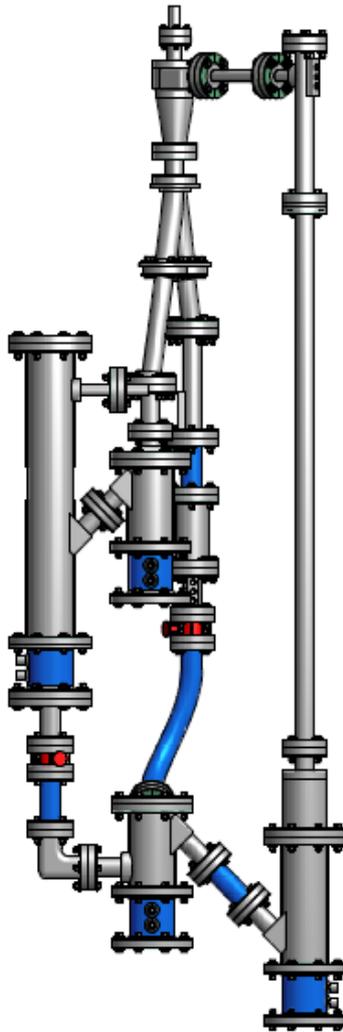
32D Hot Non-reacting Flow (TE3962 & TE3965 average)



# C2U and validation data



# C2U and validation data



## C2U Progression

1. Design and construction
2. Shakedown and modification as needed
3. Revise design drawing
4. Create system models from revised drawings

# C2U and validation data

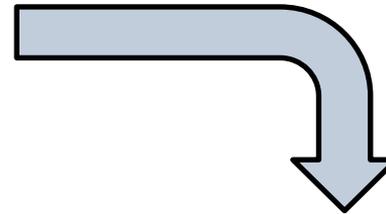
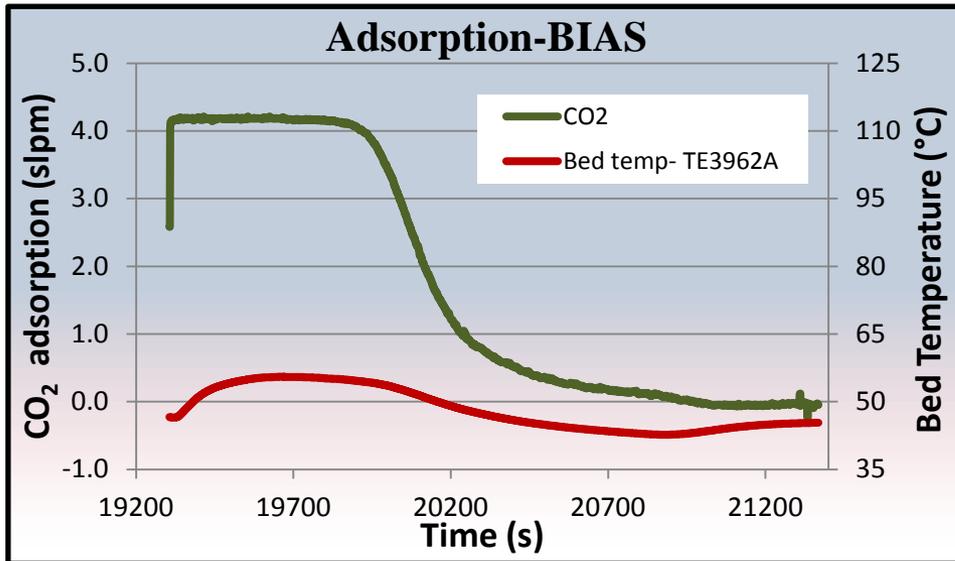
To validate individual models, three sets of tests where statistically devised and randomized

1. Cold Flow testing – hydrodynamics
2. Hot Flow testing – heat transfer, hydrodynamics
3. Reaction testing – reaction kinetics, heat transfer, hydrodynamics

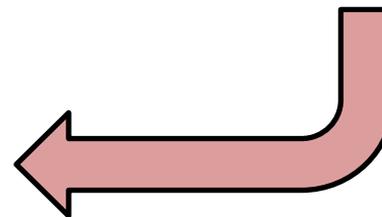
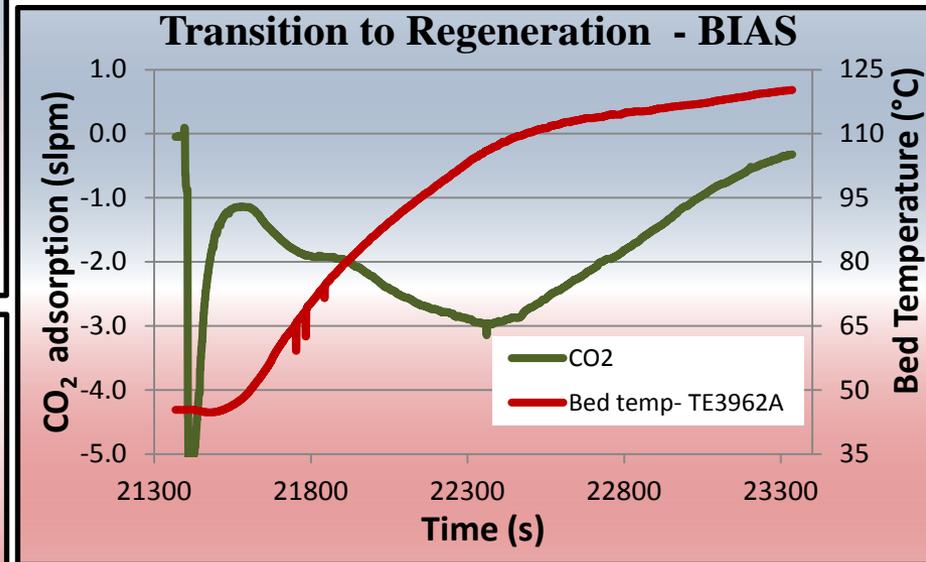
Sorbent AX			Sorbent 32D		
Cold Flow	Hot Flow		Cold Flow	Hot Flow	
Flow (SLPM)	Flow (SLPM)	Temp(°C)	Flow (SLPM)	Flow (SLPM)	Temp(°C)
49.8	16.9	70.3	19.2	21.9	60.4
15	48.8	60.2	23.6	39.5	66.7
58.9	35.4	56.3	50.3	43.8	45.3
43.7	38.6	67.4	51.7	46.3	40.3
35.7	57.1	43.5	37	36.7	65
29.8	20.8	58.4	31.9	31.9	52.1
25.1	27.1	52.4	25.9	16.4	71.1
20.4	30	42.7	51.6	26.7	47.7
54.6	48.6	49	35.4	22.7	53.4
40.3	26	77.2	25.4	52.7	78.1
38.4	54.7	65.6	59.6	58.1	57.7
23.2	41.9	74.2	33.3	51	73.4
31.6	40.6	61.8	16.1	30.6	79.1
28	52	73	27.4	52	68.8
47.8	23	47.4	21.6	48.2	55.2
18	15.8	45.3	43.3	19.3	65.7
33.4	29.5	69.7	54.9	39.3	43.5
52.6	44.4	40.5	41.4	49.4	63
46	36	50	29.6	29.7	76
56.9	19.2	76.3	47.7	56.1	42.6
15	32.2	63.8	16.1	35.2	58.8
31.6	46.1	54.4	45.9	17.1	48.5
49.8	53.3	56.8	57.5	42.3	51.6
35.7	58.7	79.4	41.4	25.8	71.9
20.4	15.8	45.3	25.4	17.1	48.5
46	38.6	67.4	33.3	30.6	79.1
25.1	36	50	21.6	58.1	57.7
43.7	58.7	79.4	29.6	51	73.4
56.9	53.3	56.8	51.7	19.3	65.7
54.6	19.2	76.3	50.3	43.8	45.3
29.8	37	31.9	31.9	48.2	54.9
38.4	43.3	47.8	19.2	54.4	75.7
47.8	19.2	27.4	37	15.1	56.4
23.2	47.8	29.6	47	34.6	47
58.9	39.7	54.9	39.7	30.2	77.6
33.4	39.7	31.9	31.9	26.7	48.3
18	47.7	59.6	28.6	19.9	44
40.3	28	59.6	28.6	28.6	50.7
28	28	28	28	25.1	74.5
52.6	23.6	23.6	23.6	22.2	76.4
				46.8	40.9
				39.2	66.1
				30.8	42.7
				44.5	69.1
				42.3	56
				53.4	72.5
				19	79.3
				27.9	73.7
				17.6	51.8
				40.5	49.4
				19	79.3
				32.8	59
				54.4	75.7
				46.8	40.9
				38.2	76.9
				24.5	67.8
				20.5	44.8
				57.5	59.3
				56.9	57.9
				35.6	41.6
					10.4

Sorbent 32D					
Cold Flow	Hot Flow		Reacting Flow		
Flow (SLPM)	Flow (SLPM)	Temp(°C)	Flow (SLPM)	Temp(°C)	CO <sub>2</sub> Conc.
19.2	21.9	60.4	51.3	62.9	18.9
23.6	39.5	66.7	40	62.3	10.8
50.3	43.8	45.3	37	68.2	14.6
51.7	46.3	40.3	27.3	72.8	15.7

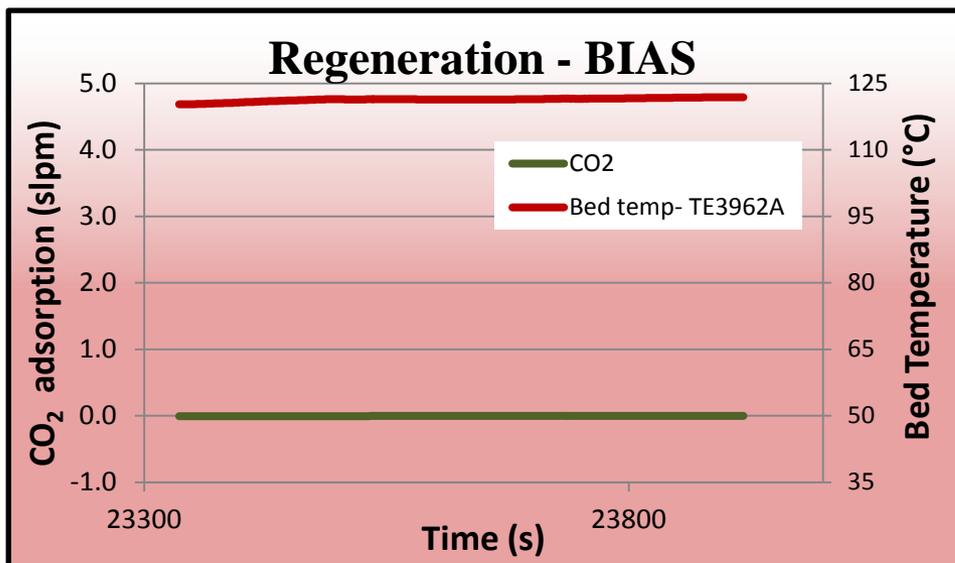
# C2U Initial testing data



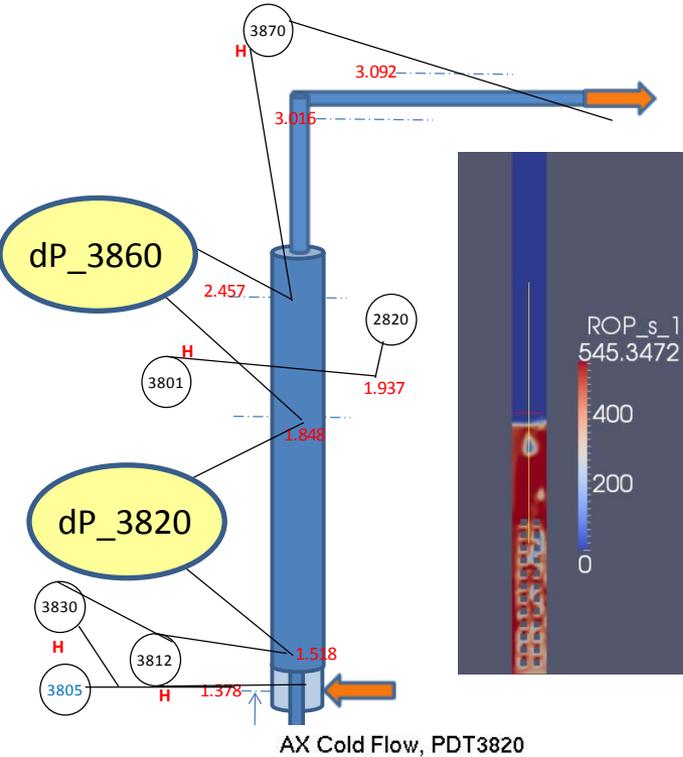
Batch test conducted in the regenerator with the BIAS sorbent



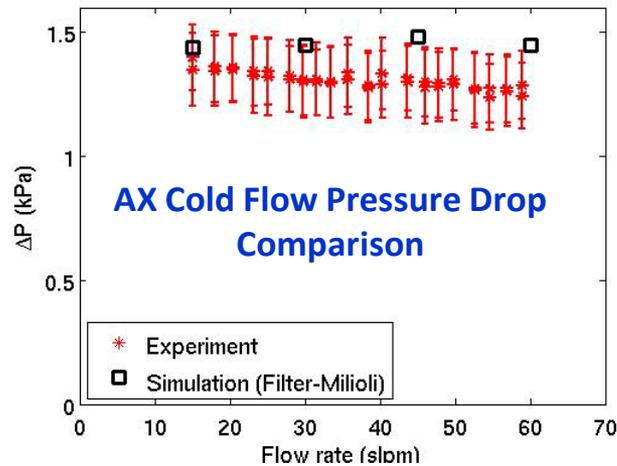
Most CO<sub>2</sub> is released before achieving the regeneration temperature



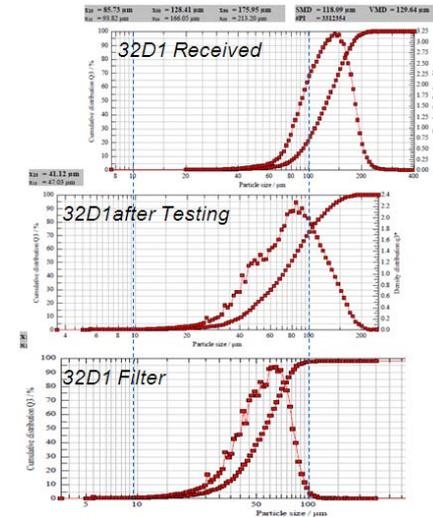
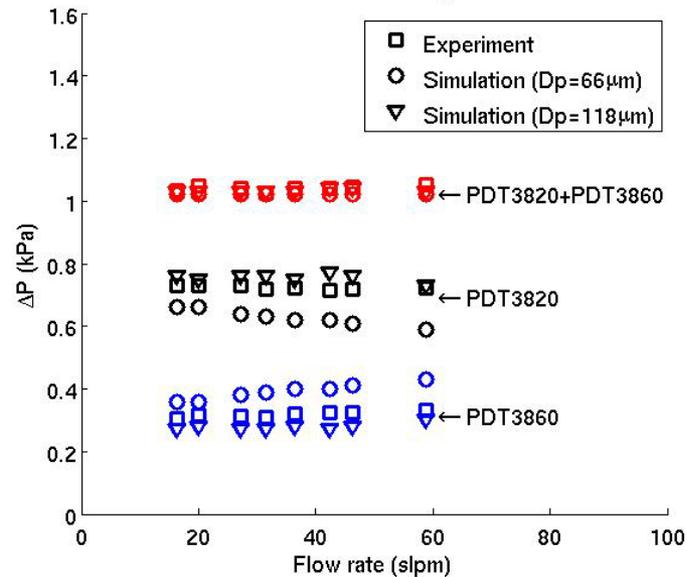
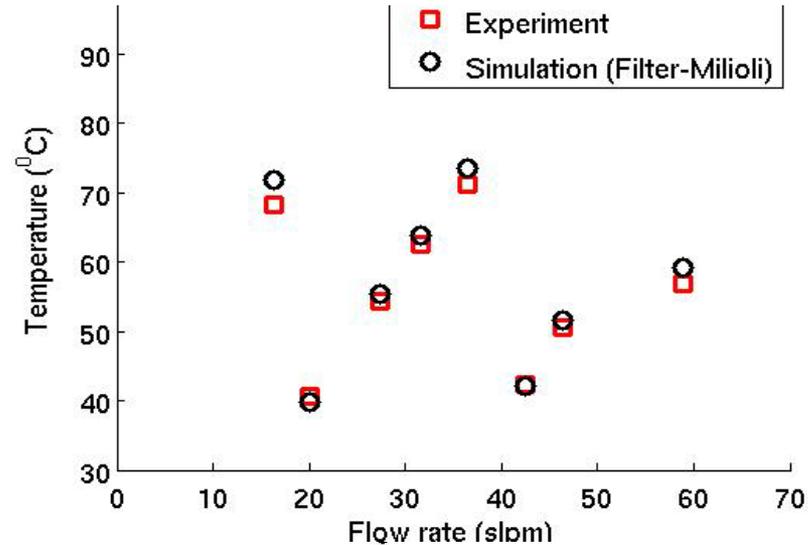
# CFD Validation with C2U non-circulating Experiments



AX Cold Flow, PDT3820

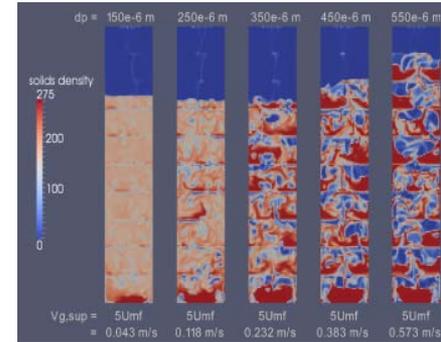
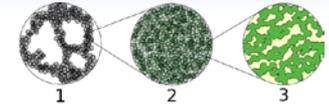
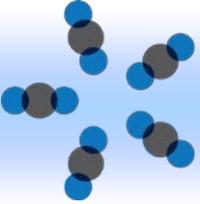


## 32D Hot Non-reactive Flow



# Conclusion

- **Coupling experimental development with simulation enables**
  - New approaches to be screened more quickly
  - Focuses development on most promising material and process conditions
- **Simulation with uncertainty quantification**
  - Focuses experimental efforts on elements with the most impact to the process/technology
- **Focused experiments for model validation enable**
  - Lower risk for scale up through quantitative confidence of model predictions



- **5-7 PM today in Elwood I**
  - Demonstration and detailed discussion of capabilities
- **Sorbents, Solvents, Membranes, Oxycombustion**
- Initial toolset released Oct. 2012
  - Four (4) companies have already licensed
  - Additional releases planned for Fall 2013, 2014, 2015
  - Final release planned for Jan. 2016

